

Memory Effects in Micro and Nanoscale Systems

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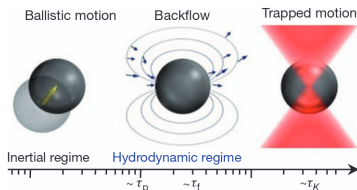
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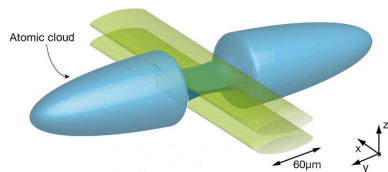
Memory Effects in Micro and Nanoscale Systems

Brownian motion

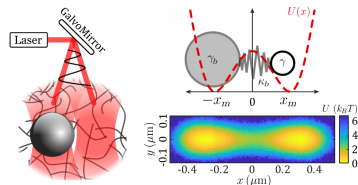


➔ T. Franosch et al., Nature **478** 85 (2011).

Finite-size reservoirs

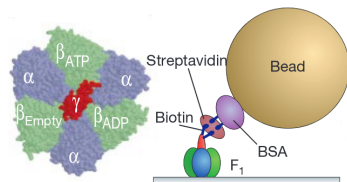


➔ S. Krinner et al., J. Phys. Condens. Matter **29** 343003 (2017).



➔ F. Ginot et al., PRL **128** 028001 (2022).

Hidden degrees of freedom



➔ R. Yasuda et al., Nature **410** 898 (2001).

Step 1: Microscopic model

$$\dot{X}_t = WX_t$$

X_t . . . state vector

W . . . microscopic generator

➤ local in time

Step 2: Coarse graining

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'}$$

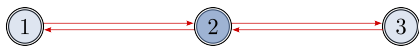
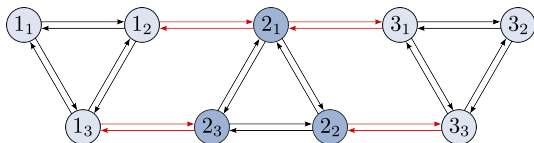
x_t . . . reduced state vector

V . . . adiabatic generator

K_t . . . memory kernel

➤ non-local in time

➤ fully systematic



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➤ non-local in time

➤ fully systematic

Step 3: Short memory approximation

$$\dot{x}_t \simeq Lx_t$$

L . . . effective generator

Adiabatic approximation

$$L^0 = V$$

Markov approximation

$$L^1 = V + \int_0^\infty dt K_t e^{-Vt}$$

➔ M. Esposito, PRE **85** 041125 (2012).

➔ G. Hummer, A. Szabo, J. Phys. Chem. B **119** 9029 (2015).

➤ local in time

➤ generally non-systematic

➤ requires sharp separation of time scales

Step 1: Microscopic model

$$\dot{X}_t = WX_t$$

X_t . . . state vector

W . . . microscopic generator

- ▶ local in time

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- ▶ non-local in time
- ▶ fully systematic

Step 3: Short memory approximation

$$\dot{x}_t \simeq Lx_t$$

L . . . effective local generator

- ▶ local in time
- ▶ generally non-systematic
- ▶ requires sharp separation of time scales

Questions

1. When and in what sense does a local approximation exist?
2. Can its error be bounded?
3. How to construct it systematically?
4. Is it unique?

Starting point

$$\dot{x}_t = Vx_t + \int_0^t dt' K_t' x_{t-t'}$$

$$x_t \in \mathbb{C}^N$$

$$V, K_t \in \mathbb{C}^{N \times N}$$

Initial condition

$$x_{t=0} = x_0$$

Weak memory condition

$$\|K_t\| \leq M e^{-kt} \quad \text{for } t \geq 0$$

Aim

$L \in \mathbb{C}^{N \times N}$ and $y_0 \in \mathbb{C}^N$ so that

$$\dot{y}_t = Ly_t \quad \text{and} \quad x_t \simeq y_t$$

Short time expansion

$$x_t = x_0 + Vx_0 \cdot t + \frac{V^2 + K_0}{2} x_0 \cdot t^2 + \mathcal{O}(t^3)$$

$$y_t = y_0 + Ly_0 \cdot t + \frac{L^2}{2} y_0 \cdot t^2 + \mathcal{O}(t^3)$$

► $x_t \neq y_t$ for short times in general

Starting point

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'}$$

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$$\dot{y}_t = Ly_t \quad \text{and} \quad x_t \simeq y_t$$

Ansatz

$$x_t = e^{Lt} A_t x_0, \quad \lim_{t \rightarrow \infty} A_t = D, \quad \det[D] \neq 0$$

$$y_t = e^{Lt} y_0, \quad y_0 = Dx_0, \quad \lim_{t \rightarrow \infty} |x_t - y_t| = 0$$

A_t . . . reduced propagator

D . . . slippage matrix

y_t . . . long time approximation

Questions

1. When does such a generator L exist?
2. Can $|x_t - y_t|$ be bounded?

Scalar Model

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'}$$

$$V = -v \leq 0$$

$$K_t = -Me^{-kt} \leq 0$$

$$x_0 = 1$$

Solution

$$x_t = \frac{k - \rho}{\eta - \rho} e^{-\rho t} - \frac{k - \eta}{\eta - \rho} e^{-\eta t}$$

$$\rho = \frac{k + v - \sqrt{(k - v)^2 - 4M}}{2}$$

$$\eta = \frac{k + v + \sqrt{(k - v)^2 - 4M}}{2}$$

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$$\eta = \frac{k + v + \sqrt{(k - v)^2 - 4M}}{2}$$

Observation

If $4M < (k - v)^2$:

$$L = -\rho, \quad A_t = e^{-Lt} x_t \rightarrow \frac{k - \rho}{\eta - \rho} = D \neq 0$$

$$y_t = e^{Lt} D, \quad |x_t - y_t| = \frac{k - \eta}{\eta - \rho} e^{-\eta t} \rightarrow 0$$

If $4M = (k - v)^2$:

$$x_t = (1 + (k - \rho)t) e^{-\rho t}$$

If $4M > (k - v)^2$:

$$x_t = \left(\cos[\omega t] + \frac{k - v}{2\omega} \sin[\omega t] \right) e^{-\delta t}$$

$$\omega = \frac{|\eta - \rho|}{2}, \quad \delta = \frac{|\eta + \rho|}{2}$$

► The condition $4M < (k - v)^2$ is necessary for the existence of an effective generator.

Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'} \quad (t \geq 0)$$

$$x_t \in \mathbb{C}^N, \quad V, K_t \in \mathbb{C}^{N \times N}, \quad x_{t=0} = x_0$$

Weak memory conditions

$$\|K_t\| \leq M e^{-kt}, \quad \|V\| = v < k, \quad \frac{4M}{(k-v)^2} \leq 1$$

$$M, k > 0, \quad v \geq 0$$

Setting

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Existence of effective generators

There exists a unique pair of generators $L, R \in \mathbb{C}^{N \times N}$ so that

$$x_t = e^{Lt} A_t x_0 = B_t e^{Rt} x_0,$$

$$\lim_{t \rightarrow \infty} A_t = \lim_{t \rightarrow \infty} B_t = D, \quad \det[D] \neq 0.$$

The slippage matrix D satisfies $LD = DR$ and

$$D = \left[1 + \int_0^\infty dt \int_0^\infty dt' e^{-Rt} K_{t+t'} e^{-Lt'} \right]^{-1}.$$

Theorem

Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'} \quad (t \geq 0)$$

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Fixed point equations

L and R are unique attractive solutions of

$$L = V + \int_0^\infty dt K_t e^{-Lt},$$

$$R = V + \int_0^\infty dt e^{-Rt} K_t,$$

$$\|L\|, \|R\| \leq \rho = \frac{k + v - \sqrt{(k-v)^2 - 4M}}{2}.$$

Theorem

Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'} \quad (t \geq 0)$$

$$x_t \in \mathbb{C}^N, \quad V, K_t \in \mathbb{C}^{N \times N}, \quad x_{t=0} = x_0$$

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$$\|K_t\| \leq M e^{-kt}, \quad \|V\| = v < k, \quad \frac{4M}{(k-v)^2} \leq 1$$

$$M, k > 0, \quad v \geq 0$$

Long time approximation

For any vector norm $|\cdot|$ consistent with the matrix norm $\|\cdot\|$ we have

$$y_t = e^{Lt} D = D e^{Rt}, \quad |x_t - y_t| \leq \frac{k - \eta}{\eta - \rho} e^{-\eta t},$$

$$\eta = \frac{k + v + \sqrt{(k - v)^2 - 4M}}{2}.$$

Existence of effective generators

There exists a unique pair of generators $L, R \in \mathbb{C}^{N \times N}$ so that

$$x_t = e^{Lt} A_t x_0 = B_t e^{Rt} x_0,$$

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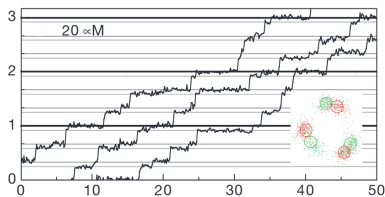
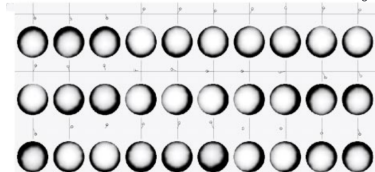
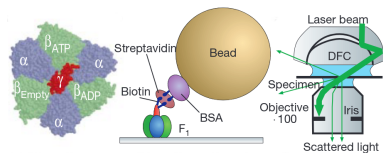
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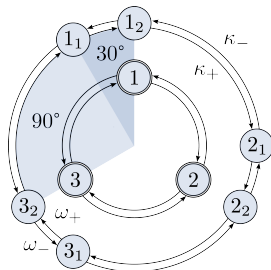
Example 1 : Molecular Motor

F₁-ATPase



➔ R. Yasuda et al., Nature **410** 898 (2001).

Model



$$x_t = [p_t^1, p_t^2, p_t^3]^T, \quad \dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'}$$

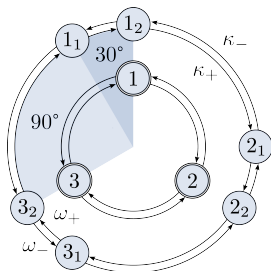
$$V = \sigma_+(S - 1) + \sigma_-(S^T - 1)$$

$$K_t = (\zeta_+(S - 1) + \zeta_-(S^T - 1))e^{-(V+\nu)t}$$

$$S = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}, \quad S^T = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix}$$

Example 1 : Molecular Motor

Model



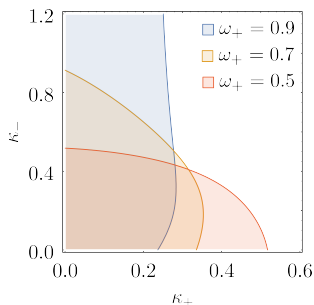
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$$V = \sigma_+(S - 1) + \sigma_-(S^T - 1)$$

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Weak memory conditions



$$\|K_t\|_2 \leq Me^{-kt},$$

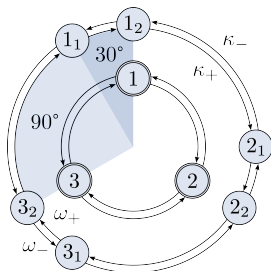
$$\|V\|_2 = v < k, \quad \frac{4M}{(k-v)^2} \leq 1$$

➤ No sharp separation of time scales required.

$$[\omega_+ + \omega_- = 1]$$

Example 1 : Molecular Motor

Model



$$x_t = [p_t^1, p_t^2, p_t^3]^T, \quad \dot{x}_t = Vx_t + \int_0^t dt' K_t' x_{t-t'}$$

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Effective generators

Fixed point equations

$$L = V + \int_0^\infty dt K_t e^{-Lt}$$

$$R = V + \int_0^\infty dt e^{-Rt} K_t$$

Solution

$$L = R = \lambda_+(S - 1) + \lambda_-(S^T - 1)$$

Slippage matrix

$$D = \left[1 + \int_0^\infty dt \int_0^\infty dt' e^{-Rt} K_{t+t'} e^{-Lt'} \right]^{-1}$$

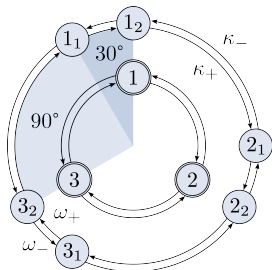
$$= \left[1 + K_0 [L + V + \kappa]^{-2} \right]^{-1}$$

Long time approximation

$$y_t = e^{Lt} D x_0$$

Example 1 : Molecular Motor

Model



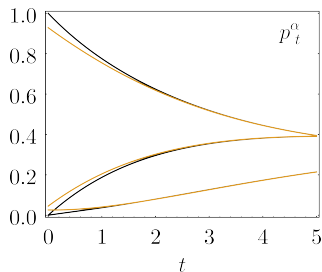
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Long time approximation



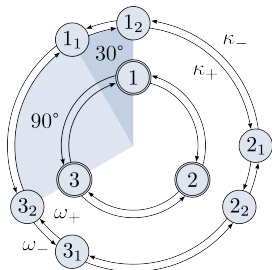
$$y_t = e^{Lt} D x_0$$

$$[\omega_+ = 0.8, \omega_- = 0.2, \kappa_+ = 0.3, \kappa_- = 0.2]$$

$$[v/k \simeq 0.42, 4M/(k - v)^2 \simeq 0.89]$$

Example 1 : Molecular Motor

Model



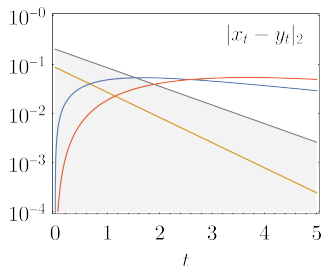
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Long time approximation



$$|x_t - y_t|_2 \leq \frac{k - \eta}{\eta - \rho} e^{-\eta t}$$

$$|x_t - y_t^M|_2, \quad y_t^M = e^{L^1 t} x_0$$

$$L^1 = V + \int_0^\infty dt K_t e^{-\nu t}$$

$$|x_t - y_t^O|_2, \quad y_t^O = e^{Vt} x_0$$

Questions

1. When and in what sense does a local approximation exist?
2. Can its error be bounded?
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Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'} \quad (t \geq 0)$$

$$\|K_t\| \leq Me^{-kt}, \quad \|V\| = v < k, \quad \frac{4M}{(k-v)^2} < 1$$

$$x_t = e^{Lt} A_t x_0$$

Memory function

$$\dot{x}_t = Lx_t + E_t x_0, \quad E_t = e^{Lt} \dot{A}_t$$

$$\dot{E}_t = K_t + E_t V + \int_0^t dt' E_{t'} K_{t-t'}, \quad E_0 = V - L$$

Rescaling

$$s = kt, \quad \bar{E}_s = E_{s/k}/k, \quad \bar{V} = V/k, \quad \bar{K}_s = K_{s/k}/M$$

$$\|\bar{K}_s\| \leq e^{-s}$$

$$\frac{d}{ds} \bar{E}_s = \varphi \bar{K}_s + \bar{E}_s \bar{V} + \varphi \int_0^s ds' \bar{E}_{s'} \bar{K}_{s-s'}, \quad \varphi = \frac{M}{k^2}$$

Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'} \quad (t \geq 0)$$

$$\|K_t\| \leq M e^{-kt}, \quad \|V\| = v < k, \quad \frac{4M}{(k-v)^2} < 1$$

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$$\|\bar{K}_s\| \leq e^{-s}$$

$$\frac{d}{ds} \bar{E}_s = \varphi \bar{K}_s + \bar{E}_s \bar{V} + \varphi \int_0^s ds' \bar{E}_{s'} \bar{K}_{s-s'}, \quad \varphi = \frac{M}{k^2}$$

Perturbation theory

Ansatz

$$\bar{E}_s = \sum_{n=1}^{\infty} \varphi^n \bar{E}_s^{(n)}, \quad \lim_{s \rightarrow \infty} \bar{E}_s^{(n)} e^{-\bar{V}s} = 0$$

Recursion relations

$$E_t^{(n)} = - \int_t^{\infty} dt' \int_0^{t'} dt'' E_{t''}^{(n-1)} K_{t'-t''} e^{V(t-t')}$$

$$E_t^{(1)} = - \int_t^{\infty} dt' K_{t'} e^{V(t-t')}$$

Approximations

$$E_t^n = \sum_{m=1}^n E_t^{(m)}, \quad L^n = V - \sum_{m=1}^n E_0^{(m)}$$

$$A_t^n = 1 + \int_0^t dt' e^{-L^n t'} E_{t'}^n$$

$$x_t^n = e^{L^n t} A_t^n x_0$$

Perturbation Theory

Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_t' x_{t-t'} \quad (t \geq 0)$$

$$\|K_t\| \leq Me^{-kt}, \quad \|V\| = v < k, \quad \frac{4M}{(k-v)^2} < 1$$

$$x_t = e^{Lt} A_t x_0$$

$$\dot{x}_t = Lx_t + E_t x_0, \quad E_t = e^{Lt} \dot{A}_t$$

Perturbation theory

$$E_t^n = \sum_{m=1}^n E_t^{(m)}, \quad L^n = V - \sum_{m=1}^n E_0^{(m)}$$

$$E_t^{(n)} = - \int_t^\infty dt' \int_0^{t'} dt'' E_{t'}^{(n-1)} K_{t'-t''} e^{v(t-t')}$$

$$E_t^{(1)} = - \int_t^\infty dt' K_{t'} e^{v(t-t')}$$

$$A_t^n = 1 + \int_0^t dt' e^{-L^n t'} E_{t'}^n$$

$$x_t^n = e^{L^n t} A_t^n x_0 \quad \varphi = \frac{M}{k^2}$$

Convergence

$$\|E_t - E_t^n\| \leq \frac{\varepsilon^n}{1-\varepsilon} \frac{M}{k-v} e^{-\mu t}$$

$$\|L - L^n\| \leq \frac{\varepsilon^n}{1-\varepsilon} \frac{M}{k-v}$$

$$\|A_t - A_t^n\| \leq \frac{\varepsilon^n}{1-\varepsilon} \alpha$$

$$|x_t - x_t^n| \leq \frac{\varepsilon^n}{1-\varepsilon} (\alpha + \beta t) |x_0| e^{\rho t}$$

$$\varepsilon = \frac{4M}{(k-v)^2} < 1 \quad \mu = \frac{k+v}{2}$$

$$\alpha = \frac{(1+3\sqrt{1-\varepsilon})\varepsilon}{8(1-\varepsilon)}$$

$$\beta = \frac{(1+\sqrt{1-\varepsilon})\varepsilon}{8\sqrt{1-\varepsilon}} (k-v)$$

$$\rho = \frac{k+v - \sqrt{(k-v)^2 - 4M}}{2}$$

Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'} \quad (t \geq 0)$$

$$\|K_t\| \leq M e^{-kt}, \quad \|V\| = v < k, \quad \frac{4M}{(k-v)^2} < 1$$

$$x_t = e^{Lt} A_t x_0$$

$$\dot{x}_t = Lx_t + E_t x_0, \quad E_t = e^{Lt} \dot{A}_t$$

Perturbation theory

$$E_t^n = \sum_{m=1}^n E_t^{(m)}, \quad L^n = V - \sum_{m=1}^n E_0^{(m)}$$

$$E_t^{(n)} = - \int_t^\infty dt' \int_0^{t'} dt'' E_{t''}^{(n-1)} K_{t'-t''} e^{V(t-t')}$$

$$E_t^{(1)} = - \int_t^\infty dt' K_{t'} e^{V(t-t')}$$

$$A_t^n = 1 + \int_0^t dt' e^{-L^n t'} E_{t'}^n$$

$$x_t^n = e^{L^n t} A_t^n x_0 \quad \varphi = \frac{M}{k^2}$$

Convergence

$$\|E_t - E_t^n\| \leq \frac{\varepsilon^n}{1-\varepsilon} \frac{M}{k-v} e^{-\mu t}$$

$$\|L - L^n\| \leq \frac{\varepsilon^n}{1-\varepsilon} \frac{M}{k-v}$$

$$\|A_t - A_t^n\| \leq \frac{\varepsilon^n}{1-\varepsilon} \alpha$$

$$|x_t - x_t^n| \leq \frac{\varepsilon^n}{1-\varepsilon} (\alpha + \beta t) |x_0| e^{\rho t}$$

$$L^0 = V$$

$$L^1 = V + \int_0^\infty dt K_t e^{-Vt}$$

► Adiabatic and Markov generators recovered in zeroth and first order.

Perturbation Theory

Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'} \quad (t \geq 0)$$

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$$\|A_t - A_t^n\| \leq \frac{\varepsilon^n}{1-\varepsilon} \alpha$$

$$|x_t - x_t^n| \leq \frac{\varepsilon^n}{1-\varepsilon} (\alpha + \beta t) |x_0| e^{\rho t}$$

Further bounds

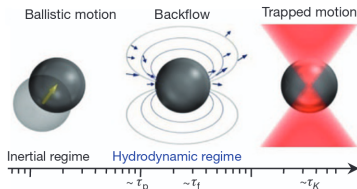
$$\text{MF:} \quad \|E_t\|, \|E_t^n\| \leq (k-\eta) e^{-\eta t}$$

$$\text{LTA:} \quad y_t^n = e^{L^n t} A_\infty^n x_0$$

$$|x_t^n - y_t^n| \leq \frac{k-\eta}{\eta-\rho} |x_0| e^{-\eta t}$$

Example 2 : Generalized Langevin Equations

Non-Markovian Brownian motion



► T. Franosch et al., Nature **478** 85 (2011).

Model

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'} + f_t$$

$$x_t = [x_t, v_t]^T, \quad \langle x_0^2 \rangle_{eq} = \langle v_0^2 \rangle_{eq} = 1$$

$$V = \begin{bmatrix} 0 & \omega \\ -\omega & 0 \end{bmatrix}, \quad K_t = \begin{bmatrix} 0 & 0 \\ 0 & -\Gamma_t \end{bmatrix}, \quad f_t = \begin{bmatrix} 0 \\ \xi_t \end{bmatrix}$$

x_t . . . particle position

v_t . . . particle velocity

ω . . . trap frequency

Γ_t . . . friction kernel

ξ_t . . . stochastic force

$$\Gamma_t = Me^{-kt},$$

$$M > 0, \quad k = 1$$

Example 2 : Generalized Langevin Equations

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$$\Gamma_t = M e^{-kt},$$

$$M > 0, \quad k = 1$$

Weak memory conditions

$$v = \|V\|_\infty = \omega < 1, \quad \frac{4M}{(\omega - 1)^2} < 1$$

Equilibrium correlation matrix

$$Z_t = \langle x_t x_0^T \rangle_{eq} [\langle x_0 x_0^T \rangle_{eq}]^{-1}$$

$$\dot{Z}_t = VZ_t + \int_0^t dt' K_{t'} Z_{t-t'}, \quad Z_0 = 1$$

$$Z_t = e^{Lt} A_t = e^{Lt} + \int_0^t dt' e^{L(t-t')} E_{t'}$$

General solution

$$x_t = Z_t x_0 + \int_0^t dt' Z_{t'} f_{t-t'}$$

Example 2 : Generalized Langevin Equations

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$$Z_t = e^{Lt} A_t = e^{Lt} + \int_0^t dt' e^{L(t-t')} E_{t'}$$

Perturbation theory

$$E_t^1 = H_{11} e^{-t}, \quad L^1 = V - H_{11}$$

$$E_t^2 = (H_{21} + H_{22t}) e^{-t}, \quad L^2 = V - H_{21}$$

$$H_{11} = \frac{M}{1 + \omega^2} \begin{bmatrix} 0 & 0 \\ \omega & 1 \end{bmatrix}$$

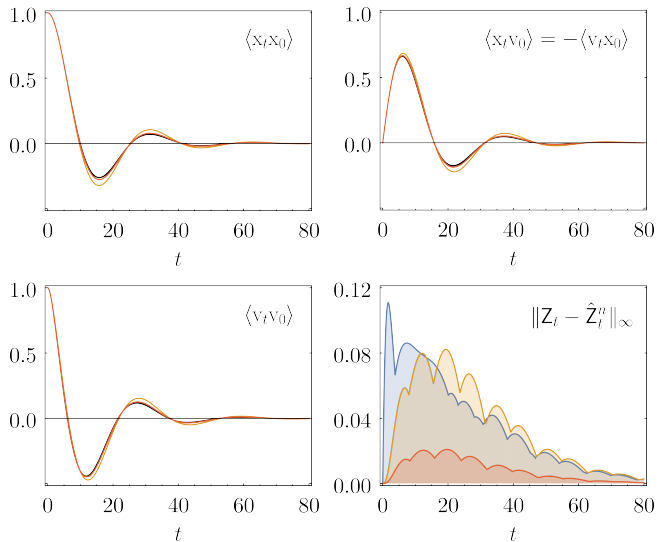
$$H_{21} = H_{11} + \frac{M^2}{(1 + \omega^2)^3} \begin{bmatrix} 0 & 0 \\ 2\omega & 1 - \omega^2 \end{bmatrix}$$

$$H_{22} = \frac{M}{1 + \omega^2} H_{11}$$

$$Z_t^M = e^{L^1 t}$$

$$Z_t^n = e^{L^n t} A_t^n = e^{L^n t} + \int_0^t dt' e^{L^n(t-t')} E_{t'}^n$$

Example 2 : Generalized Langevin Equations



$$Z_t^M = e^{L^M t}$$

$$Z_t^1 = e^{L^1 t} A_t^1$$

$$Z_t^2 = e^{L^2 t} A_t^2$$

[$\omega = 0.2$, $M = 0.15$, $k = 1$, $4M/(1 - \omega^2) \simeq 0.94$]

Questions

1. When and in what sense does a local approximation exist?
2. Can its error be bounded?
3. How to construct it systematically?
4. Is it unique?

Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'} \quad (t \geq 0)$$

$$\|K_t\| \leq Me^{-kt}, \quad \|V\| = v < k, \quad \frac{4M}{(k-v)^2} < 1$$

$$x_t = Z_t x_0$$

Two-sided factorization

$$\lim_{t \rightarrow \infty} e^{-Lt} Z_t = \lim_{t \rightarrow \infty} Z_t e^{-Rt} \stackrel{!}{=} D, \quad \det[D] \stackrel{!}{\neq} 0$$

$\Rightarrow L, R$ uniquely determined.

One-sided factorization

$$\lim_{t \rightarrow \infty} e^{-L't} Z_t \stackrel{!}{=} D', \quad \det[D'] \stackrel{!}{\neq} 0$$

$$\Rightarrow L' = SLS^{-1}, \quad D' = SD$$

$$\lim_{t \rightarrow \infty} e^{Lt} S^{-1} e^{-Lt} = 1, \quad \det[S] \neq 0$$

Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_t' x_{t-t'} \quad (t \geq 0)$$

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$$\lim_{t \rightarrow \infty} e^{Lt} S^{-1} e^{-L't} = 1, \quad \det[S] \neq 0$$

Fixed point equations

$$L \stackrel{!}{=} V + \int_0^\infty dt K_t e^{-Lt}, \quad \|L\| \stackrel{!}{\leq} \rho$$

$$R \stackrel{!}{=} V + \int_0^\infty dt e^{-Rt} K_t, \quad \|R\| \stackrel{!}{\leq} \rho$$

$\Rightarrow L, R$ uniquely determined.

Memory functions

$$\dot{Z}_t = LZ_t + E_t = Z_t R + F_t$$

$$\|E_t\|, \|F_t\| \leq (k - \eta) e^{-\eta t}$$

$\Rightarrow L, R$ uniquely determined.

For any $L' \neq L, R' \neq R$ and $\sigma > \rho$:

$$\dot{Z}_t = L'Z_t + E'_t = Z_t R' + F'_t$$

$$\lim_{t \rightarrow \infty} \|E'_t\| e^{\sigma t} = \lim_{t \rightarrow \infty} \|F'_t\| e^{\sigma t} = \infty \quad (\rho < \eta)$$

Setting

$$\dot{x}_t = Vx_t + \int_0^t dt' K_t' x_{t-t'} \quad (t \geq 0)$$

$$\|K_t\| \leq M e^{-kt}, \quad \|V\| = v < k, \quad \frac{4M}{(k-v)^2} < 1$$

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Fixed point equations

$$L \stackrel{!}{=} V + \int_0^\infty dt K_t e^{-Lt}, \quad \|L\| \stackrel{!}{\leq} \rho$$

$$R \stackrel{!}{=} V + \int_0^\infty dt e^{-Rt} K_t, \quad \|R\| \stackrel{!}{\leq} \rho$$

\Rightarrow L, R uniquely determined.

Memory functions

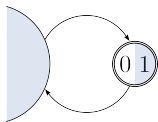
$$\dot{Z}_t = LZ_t + E_t = Z_t R + F_t$$

$$\|E_t\|, \|F_t\| \leq (k - \eta) e^{-\eta t}$$

\Rightarrow L, R uniquely determined.

- \blacktriangleright The proper generators lead to the fastest decaying memory functions.

Example 3 : Semi-Markov Jump Process



Waiting time distribution ($\kappa > 4$)

$$\psi_t = \frac{2\gamma\kappa}{\sqrt{\kappa(\kappa-2)}} \sinh \left[\sqrt{\kappa(\kappa-2)}\gamma t \right] e^{-\kappa\gamma t}$$

Master equation

$$\dot{x}_t = \int_0^t dt' K_{t'} x_{t-t'}, \quad x_t = [p_t^0, p_t^1]^T$$

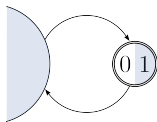
$$K_t = M e^{-kt} H_1, \quad M = 4\kappa\gamma^2, \quad k = 2\kappa\gamma$$

$$H_1 = \frac{1}{2} \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}, \quad \|H_1\|_1 = 1$$

$$\rho = \gamma \left(\kappa - \sqrt{\kappa(\kappa-4)} \right)$$

$$\eta = \gamma \left(\kappa + \sqrt{\kappa(\kappa-4)} \right) > \rho$$

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Master equation

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$$H_1 = \frac{1}{2} \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}, \quad \|H_1\|_1 = 1$$

$$\rho = \gamma \left(\kappa - \sqrt{\kappa(\kappa-4)} \right)$$

$$\eta = \gamma \left(\kappa + \sqrt{\kappa(\kappa-4)} \right) > \rho$$

Modified generator

$$L = \rho H_1$$

$$\lim_{t \rightarrow \infty} e^{Lt} S^{-1} e^{-Lt} = 1 \Rightarrow S = S_\alpha = 1 + \alpha H_2$$

$$H_1 = \frac{1}{2} \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}, \quad \alpha \in \mathbb{C}$$

$$L_\alpha = S_\alpha L S_\alpha^{-1} = \rho H_1 - \alpha \rho H_2$$

Modified memory function

$$E_{\alpha,t} = -\rho e^{-\eta t} H_1 + \frac{\alpha \rho}{\eta - \rho} \left(\eta e^{-\rho t} - \rho e^{-\eta t} \right) H_2$$

For $\alpha = 0$:

$$\|E_t\| = \rho e^{-\eta t} = (k - \eta) e^{-\eta t}$$

For $\alpha \neq 0$:

$$\|E_{\alpha,t}\| = \rho e^{-\eta t} + \frac{\alpha \rho}{\eta - \rho} \left(\eta e^{-\rho t} - \rho e^{-\eta t} \right)$$

Questions

1. When and in what sense does a local approximation exist?
2. Can its error be bounded?
3. How to construct it systematically?
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Time Convolutionless Method

TCL approach

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'}$$

$$\rightarrow \dot{x}_t = L_t^{\text{TCL}} x_t$$

Present approach

$$\rightarrow \dot{x}_t = L + E_t x_0$$

Under WM conditions:

$$\lim_{t \rightarrow \infty} L_t^{\text{TCL}} = \lim_{t \rightarrow \infty} \dot{Z}_t Z_t^{-1} = L$$

➔ M. Tokuyama and H. Mori., Prog. Theor. Phys. **55** 411 (1976).

➔ F. Shibata and T. Arimitsu, J. Phys. Soc. Jpn. **49** 891 (1980).

Slippage matrix

$$e^{L^1 t} x_0 \rightarrow e^{L^1 t} D x_0$$

➔ F. Haake and M. Lewenstein, PRA **28** 3606 (1983).

➔ P. Gaspard and M. Nagaoka, J. Chem. Phys. **111** 5668 (1999).

Fixed point equations

$$L_t^{\text{TCL}} = VL_t^{\text{TCL}} + \int_0^t dt' K_{t'} \mathcal{T} \exp \left[- \int_0^{t'} dt'' L_{t-t''}^{\text{TCL}} \right]$$

If $L_t^{\text{TCL}} \rightarrow L$ for $t \rightarrow \infty$:

$$\rightarrow L = VL + \int_0^\infty dt K_t e^{-Lt}$$

➔ K. Nestmann et al., PRX **11** 021041 (2021).

➔ K. Nestmann, M. R. Wegewijs, PRB **104** 155407 (2021).

➔ V. Bruch et al., SciPost Phys. **11** 053 (2021).

Memory expansions

$$L = \sum_{n=0}^{\infty} L^{(n)}$$

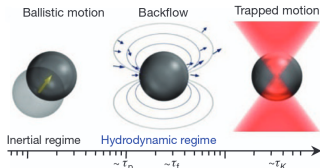
Assuming convergence:

$$\rightarrow L = VL + \int_0^\infty dt K_t e^{-Lt}$$

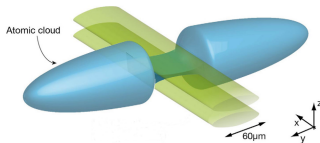
➔ L. D. Contreras-Pulido et al., PRB **85** 075301 (2012).

➔ C. Karlewski and M. Marthaler, PRB **90** 104302 (2014).

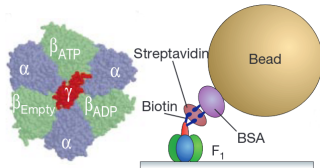
Summary



➔ T. Franosch et al., Nature **478** 85 (2011).



➔ S. Krinner et al., JP CM **29** 343003 (2017).



➔ R. Yasuda et al., Nature **410** 898 (2001).

Key results

$$\dot{x}_t = Vx_t + \int_0^t dt' K_{t'} x_{t-t'}$$

Weak memory conditions

$$\|K_t\| \leq Me^{-kt}, \quad \|V\| = v < k, \quad \frac{4M}{(k-v)^2} < 1$$

Factorization

$$x_t = e^{Lt} A_t x_0, \quad \lim_{t \rightarrow \infty} A_t = D, \quad \det[D] \neq 0$$

Long time approximation

$$y_t = e^{Lt} D x_0, \quad |y_t - x_t| \leq \frac{k - \eta}{\eta - \rho} e^{-\eta t}$$

Quasi local time evolution equation

$$\dot{x}_t = Lx_t + E_t x_0, \quad \|E_t\| \leq (k - \eta) e^{-\eta t}$$

Convergent perturbation scheme

$$\bar{X} = \sum_{n=0}^{\infty} \varphi^n \bar{X}^{(n)}, \quad X = L, E_t, \quad \varphi = \frac{M}{k^2} A_t$$