

Non-Linear Thermal Conductance and Noise Dependence

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Welles A. M. Morgado

PUC-Rio



- Silvio M. Duarte Queirós (CBPF)
- Diogo O. Soares-Pinto (IFSC-USP)
- Celia Anteneodo (PUC-Rio)
- Michael M. Cândido (aluno de doutorado, PUC-Rio)
- Pierre Soares (aluno de doutorado, PUC-Rio)

References:

PRE **79** 051116 (2009)

JSTAT P06010 (2011)

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- Simple small models can be quite illustrative about the way biological and molecular systems behave.
- We show how the couplings between the model's non-linearities and non-equilibrium properties of thermal noise lead to new properties of transport coefficients.

Why are we interested in this problem?

- Under certain conditions, there might be non-equilibrium energy feeding into a system (mostly small), i.e, non-gaussian noises will act upon these systems. For instance, Poisson noise.
- We might be able to solve small system models with arbitrary noise properties.
- What happens when non-linearities couple with the non-gaussian (higher cumulants) properties of noise?

Equilibrium noise

- An equilibrium thermal reservoir interacts with a system, say a Brownian particle, by an effective interaction that may be well represented by white gaussian noise.
- That form of noise ensures the Boltzmann-Gibbs equilibrium distribution for the system.
- Only the mean and variance of the noise exists.
- For non-equilibrium noise, higher order cumulants may exist.

Non-Equilibrium Stationary State - NESS

- A system in out-of equilibrium conditions might reach a NESS.
- This can be done by submitting the system to contact with distinct temperature reservoirs, or by submitting the system to a non-equilibrium noise, e.g. Poisson noise.
- We shall see that the distinction above is less than it seems.

Poisson noise & Brownian particles

A Brownian particle under Poisson noise is a paradigmatic NESS for $t \rightarrow \infty$.

The model is

$$\dot{x}(t) = v(t),$$

$$M \dot{v}(t) = -k_0 x(t) - \gamma v(t) + \eta(t),$$

where η is a Poisson noise,

$$\eta(t) = \sum_{\ell} \Phi(t) \delta(t - t_{\ell}).$$

It is interesting to chose a variable rate

$$\lambda(t) = \lambda_0 [1 + A \cos(\omega t)],$$

Noise Cumulants

- Poisson noise has infinite number of cumulants:

$$\langle \eta_{i_1}(t_1) \eta_{i_2}(t_2) \dots \eta_{i_n}(t_n) \rangle_c = \overline{\Phi^n} \lambda(t_1) \delta_{i_1 i_2} \dots \delta_{i_{n-1} i_n} \\ \times \delta(t_1 - t_2) \dots \delta(t_{n-1} - t_n).$$

- The general Laplace transform for noise reads:

$$\langle \tilde{\eta}_1(s_1) \dots \tilde{\eta}_n \rangle_c = \frac{A_n}{s_1 + \dots + s_n},$$

where A_n is the n-th order cumulant

- Energy injection, dissipation, and the received heat ($x_0 = v_0 = 0$):

$$Q = \int_0^\tau [\eta(t) v(t) - \gamma v(t)^2] dt = \frac{1}{2} M v(t)^2 \Big|_0^\tau + \frac{1}{2} k_0 x(t)^2 \Big|_0^\tau.$$

The time averaged exact NESS energy:

Then finally, we write the contributions for the injection of energy as

$$J_{IT0}(\tau) = \left(\frac{\lambda_0 \bar{\Phi}^2}{M} + \frac{A^2 \omega^2 \theta \lambda_0^2 \bar{\Phi}^2}{M [\omega^2 (\omega^2 - 2\omega_0^2 + 4\theta^2) + \omega_0^4]} \right) \tau,$$

and

$$J_{IT,osc}(\tau) = \frac{\lambda_0 \bar{\Phi}^2 A \sin(\omega \tau)}{M \omega} + \frac{A [A (\omega_0^2 - \omega^2) \cos(2\omega \tau) + 4\omega_0^2 \cos(\omega \tau)] \lambda_0^2 \bar{\Phi}^2}{4M [\omega^2 (\omega^2 - 2\omega_0^2 + 4\theta^2) + \omega_0^4]} + \frac{A [2\omega \theta (A \sin(2\omega \tau) + 4 \sin(\omega \tau)) - 4\omega^2 \cos(\omega \tau)] \lambda_0^2 \bar{\Phi}^2}{4M [\omega^2 (\omega^2 - 2\omega_0^2 + 4\theta^2) + \omega_0^4]}, \quad (47)$$

and

$$J_{ITc} = \frac{\lambda_0^2 \bar{\Phi}^2}{M \omega_0^2} + \frac{A \lambda_0^2 \bar{\Phi}^2 (\omega_0^2 - \omega^2)}{M [\omega^2 (\omega^2 - 2\omega_0^2 + 4\theta^2) + \omega_0^4]} + \frac{[\omega^4 (\omega^2 - 12\theta^2) + \omega_0^4 (3\omega_0^4 - 5\omega_0^2 \omega^2 + \omega^4 - 4\theta^2)] A^2 \lambda_0^2 \bar{\Phi}^2}{4M [\omega^2 (\omega^2 - 2\omega_0^2 + 4\theta^2) + \omega_0^4]^2}. \quad (48)$$

For the total energy flux $J_E = J_{IT} + J_{DT}$ it can be easily seen that the linear term on τ cancels out, while the constant term becomes exactly the average energy of equation (40). The non-oscillating part of the energy is

$$J_E^{no} = \frac{\bar{\Phi}^2 \lambda_0 (2M\omega_0^2 + \lambda_0 \gamma)}{2\gamma M \omega_0^2} + \frac{(\omega^2 + \omega_0^2) A^2 \bar{\Phi}^2 \lambda_0^2}{4M [\omega^2 (\omega^2 - 2\omega_0^2 + 4\theta^2) + \omega_0^4]}, \quad (49)$$

Ressonating Energy Injection 2

Injected - Dissipated Energy = Heat

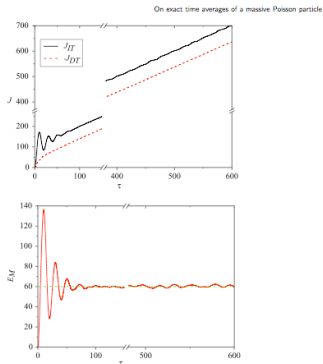


Figure 4. Upper panel: total injected and (symmetric) dissipated power, $J_{IT}(-J_{DT})$ versus time, τ , according to the definitions in the legend. Lower panel: total energy, E_M , versus time, τ . The dashed (green) line represents the asymptotic limit given by equation (40). In both cases we have used the following values: $M = 10$, $k_0 = 1$, $\gamma = 1$, $\Phi = 1$, $\lambda_0 = 10$ and $A = 0$. Note: the difference between the values of J_{IT} and $-J_{DT}$ in the upper panel are exactly equal to the total energy, E_M , which is shown in the lower panel and equals the theoretical value given by equation (40) as well.

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Ressonating Energy Injection 3

The oscillation amplitude is modulated by the resonance between the natural frequency $\omega_0 = \sqrt{\frac{k}{m}}$ and the poisson frequency ω .

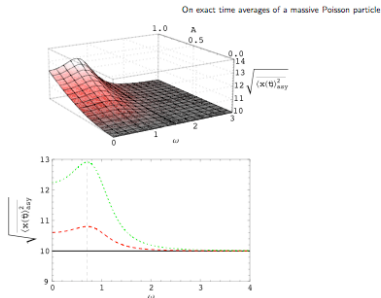


Figure 3. Upper panel: $\sqrt{\langle x(t)^2 \rangle_{avg}}$ versus amplitude of the heterogeneous part, A , and frequency, ω , with $M = 1$, $\gamma = 1$, $k_0 = 1$, $\lambda_0 = 10$, $\bar{\phi} = 1$. The maximum, characterizing a stochastic resonance phenomenon, occurs at $\omega_{res} = \sqrt{\omega_0^2 - 2\theta^2}$ that defines the grey dashed line in the lower panel containing the plane cuts for $A = 0$ (black line), $A = \frac{1}{2}$ (dashed red line) and $A = 1$ (dotted green line). For this case the maximum occurs at $1/\sqrt{2}$.

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Stationary Distributions

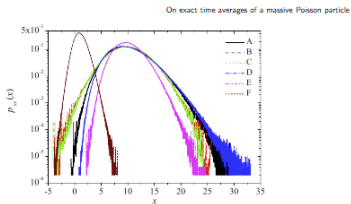


Figure 1. Numerically obtained probability density function $p_{ss}(x)$ versus position x for various cases with $\lambda_0 = 10$, $\phi = 1$ and the noise defined by equation (4) with $\omega = \pi$. Following the legend in the figure we have the respective cases, A: $M = 1, k_0 = 1, \gamma = 1, A = 0$, B: $M = 10, k_0 = 1, \gamma = 1, A = 0$, C: $M = 10, k_0 = 1, \gamma = 1, A = 1/2$, D: $M = 0.1, k_0 = 1, \gamma = 1, A = 0$, E: $M = 1, k_0 = 1, \gamma = 2, A = 0$ and F: $M = 1, k_0 = 10, \gamma = 1, A = 0$.

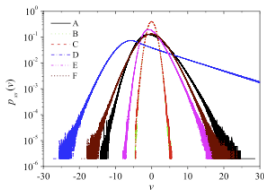


Figure 2. Numerically obtained probability density function $p_{ss}(v)$ versus scalar velocity v for the same parameter sets of figure 1.

Linear Poisson model and energy

- We did not explore fully the Poisson model due to its linearity.
- We could not couple the energy with the higher order Poisson cumulants.
- The stationary state is a NESS one: the probability distribution is not Boltzmann-Gibbs, hence NE.

- We start from (1D Brownian Particle)

$$p(x, v, t) = \langle \delta(x - x(t)) \delta(v - v(t)) \rangle,$$

where $x(t)$ and $v(t)$ are solutions for the equations of motion given a realization of the noise.

- Time averaging:

$$p^{ss}(x, v) = \lim_{s \rightarrow 0} s \int_0^{\infty} dt e^{-st} \langle \delta(x - x(t)) \delta(v - v(t)) \rangle.$$

After a straightforward calculation...

$$\begin{aligned} p^{ss}(x, v) &= \lim_{s \rightarrow \infty} \lim_{\epsilon \rightarrow \infty} \int_{-\infty}^{\infty} \frac{dQ}{2\pi} e^{iQx} \int_{-\infty}^{\infty} \frac{dP}{2\pi} e^{iPv} \times \\ &\times \sum_{n=0}^{\infty} \sum_{l=0}^{\infty} \frac{(-iQ)^n}{n!} \frac{(-iP)^l}{l!} \times \\ &\times \int_{-\infty}^{\infty} \frac{dq_1}{2\pi} \cdots \int_{-\infty}^{\infty} \frac{dq_n}{2\pi} \int_{-\infty}^{\infty} \frac{dp_1}{2\pi} \cdots \int_{-\infty}^{\infty} \frac{dp_l}{2\pi} \times \\ &\times \frac{s}{s - (iq_1 + \epsilon + \dots + iq_n + \epsilon + ip_1 + \epsilon + \dots + ip_l + \epsilon)} \times \\ &\times \langle \tilde{x}(iq_1 + \epsilon) \dots \tilde{x}(iq_n + \epsilon) \tilde{v}(ip_1 + \epsilon) \dots \tilde{v}(ip_l + \epsilon) \rangle. \end{aligned}$$

Cumulant-energy coupling

- Due to the linearity of the model ($\tilde{v}(s) = s \tilde{x}$)

$$x, v(s) = \frac{\tilde{\eta}(s)}{R(s)}.$$

- The exchanged heat Q is quadratic in $\tilde{\eta}$, hence

$$Q \propto \langle \tilde{\eta}(s_1) \tilde{\eta}(s_2) \rangle.$$

- We see that heat can only be a function of the variance, and average, of the noise

$$\langle \tilde{\eta}(s_1) \tilde{\eta}(s_2) \rangle = \langle \tilde{\eta}(s_1) \tilde{\eta}(s_2) \rangle_c + \langle \tilde{\eta}(s_1) \rangle_c \langle \tilde{\eta}(s_2) \rangle_c$$

Non-linear Interactions

- For a model with non-linear springs (non-linear force $= -k_3 x^3$) we can write:

$$R(s) \tilde{x}(s) + k_3 \int_{-\infty}^{\infty} \frac{dq_1}{2\pi} \frac{dq_2}{2\pi} \frac{dq_3}{2\pi} \frac{\tilde{x}(iq_1)\tilde{x}(iq_2)\tilde{x}(iq_3)}{s - i(q_1 + q_2 + q_3)} = \tilde{\eta}(s).$$

- The recurrence relation above shows that we can now couple the non-linearity with the higher order cumulants.
- In first order on k_3 :

$$\langle \tilde{x}(s_1) \tilde{x}(s_2) \rangle \rightarrow k_3 \int_{-\infty}^{\infty} \frac{dq_1}{2\pi} \frac{dq_2}{2\pi} \frac{dq_3}{2\pi} \frac{\langle \tilde{x}(s_1) \tilde{x}(iq_1)\tilde{x}(iq_2)\tilde{x}(iq_3) \rangle}{s_1 - i(q_1 + q_2 + q_3)}$$

where we can express $\langle \tilde{x}(s_1) \tilde{x}(iq_1) \tilde{x}(iq_2) \tilde{x}(iq_3) \rangle$ as noisy combinations of averages, variances, skewness and kurtosis!

Higher order cumulants as distinct reservoirs

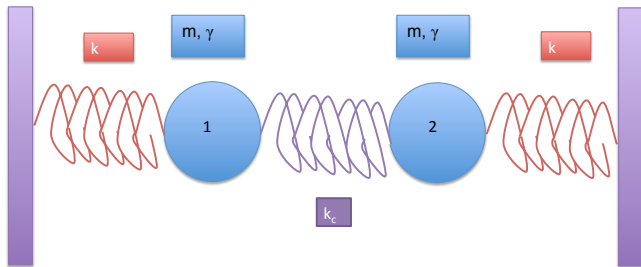
- Assuming the odd cumulants to be zero, we can express:

$$\begin{aligned}\langle \tilde{x}(s_1) \tilde{x}(iq_1) \tilde{x}(iq_2) \tilde{x}(iq_3) \rangle &= \langle \tilde{x}(s_1) \tilde{x}(iq_1) \tilde{x}(iq_2) \tilde{x}(iq_3) \rangle_c + \\ &+ \langle \tilde{x}(s_1) \tilde{x}(iq_1) \rangle_c \langle \tilde{x}(iq_2) \tilde{x}(iq_3) \rangle_c + \\ &+ \langle \tilde{x}(s_1) \tilde{x}(iq_2) \rangle_c \langle \tilde{x}(iq_1) \tilde{x}(iq_3) \rangle_c + \\ &+ \langle \tilde{x}(s_1) \tilde{x}(iq_3) \rangle_c \langle \tilde{x}(iq_2) \tilde{x}(iq_1) \rangle_c.\end{aligned}$$

- The received heat term has the (usual) contributions from the gaussian part (variance), but also the extra (unusual) kurtosis contribution.
- Thus, variance and kurtosis act as if distinct energy reservoirs are coupled to the system.

Small systems: our model

Two connected Brownian particles; non-linear springs; two thermal reservoirs.



Langevin Equation

Equations to be solved

$$m \frac{dv_i(t)}{dt} = -k x_i(t) - \gamma v_i(t) - \sum_{l=1}^2 k_{2l-1} [x_i(t) - x_j(t)]^{2l-1} + \eta_i(t)$$

with

$$\frac{dx_i(t)}{dt} = v_i(t)$$

where $(i, j) \in \{1, 2\}$ and k_1 and k_3 are the linear and non-linear coupling constants, respectively. The system is decoupled (linear) for $k_{1,3} = 0$.

Heat flux between particles

- We define the energy flux as

$$j_{12} = \frac{1}{2}(P_{21} - P_{12}) = \frac{1}{2}(F_{21} \cdot v_2 - F_{12} \cdot v_1),$$

where $F_{21} = -k_1(x_2 - x_1) - k_3(x_2 - x_1)^3$.

- We have

$$j_{12}(t) \equiv - \sum_{l=1}^2 \frac{k_{2l-1}}{2} [x_1(t) - x_2(t)]^{2l-1} [v_1(t) + v_2(t)].$$

We are finally in the position to compute the thermal conductance,

$$\begin{aligned}\kappa &\equiv -\frac{\partial}{\partial \Delta T} \langle j_{12} \rangle_{\Delta T}, \\ \kappa &= -\frac{\overline{\langle j_{12} \rangle}}{T_1 - T_2}, \quad (\text{for small } T_1 - T_2)\end{aligned}$$

Resorting to the single particle results and the equipartition theorem, we relate the cumulant features of the noise and the proper temperature, T_i , namely, $\mathcal{A}_i(2) = 2\gamma T_i$, yielding the thermal conductance, $\kappa = \kappa^{(0)} + \kappa^{(1)} + \mathcal{O}(k_3^2)$

Conductance: linear systems and general types of noises

- For linear systems, the higher order cumulants do not contribute. thus, the time averaged heat flux is **simple** and is given by

$$\overline{\langle j_{12}^{(B)} \rangle} = \overline{\langle j_{12}^{(0)} \rangle} = -\frac{k_1^2}{4} \frac{(\mathcal{A}_1(2) - \mathcal{A}_2(2))}{m k_1^2 + \gamma^2 (k + k_1)},$$

where $\mathcal{A}_2(2) - \mathcal{A}_1(2) = 2\gamma \Delta T$.

- The linear contribution for the conductance is thus

$$\kappa_0 = \frac{k_1^2 \gamma}{2 (m k_1^2 + \gamma^2 (k + k_1))},$$

Conductance: non-linear systems and gaussian noises

- The reservoirs are equilibrium ones
- First order on k_3

$$\overline{\langle j_{12}^{(1)} \rangle} = -\frac{3}{8} \gamma k_1 k_3 \frac{(2k + k_1) [\mathcal{A}_1(2)^2 - \mathcal{A}_2(2)^2]}{(k + 2k_1) [\gamma^2 (k + k_1) + m k_1^2]^2},$$

where $\mathcal{A}_1(2)^2 - \mathcal{A}_2(2)^2 = -4 \gamma^2 \Delta T (T_1 + T_2)$.

- Thus, for the non-linear model and gaussian noise we have the conductance:

$$\kappa_{0,1} = \frac{k_1^2 \gamma}{2 (m k_1^2 + \gamma^2 (k + k_1))} + \frac{3}{2} k_1 k_3 \frac{(2k + k_1) \gamma^3 (T_1 + T_2)}{(k + 2k_1) [\gamma^2 (k + k_1) + m k_1^2]^2}.$$

Conductance: non-linear systems and general noises

- The reservoirs are non-equilibrium ones
- The variance contribution still holds. however there is a new flux originating from the kurtosis in first order on k_3

$$\overline{\langle J_{12}^{(1, Poisson)} \rangle} = -\frac{27}{2} \frac{\gamma^2 k_1 k_3 \mathcal{N}}{\lambda \mathcal{D}} [\mathcal{A}_1(2)^2 - \mathcal{A}_2(2)^2],$$

where $\mathcal{A}_1(2)^2 - \mathcal{A}_2(2)^2 = -4 \gamma^2 \Delta T (T_1 + T_2)$, and

$$\mathcal{N} \equiv \gamma^2 (5 k + 3 k_1) + m (3 k_1^2 + 4 k^2 + 11 k k_1),$$

$$\mathcal{D} \equiv [\gamma^2 (k + k_1)] \left[m (4 k + 9 k_1)^2 + 6 \gamma^2 (2 k + 3 k_1) \right] \times \\ \times [3 \gamma^4 + m^2 k_1^2 + 4 m \gamma^2 (k + k_1)]$$

Conductance: non-linear systems and general noises

Thus, for the non-linear model and general (Poisson) noise we have the conductance:

$$\kappa_{1,Poisson} = 54 \frac{\gamma^4 k_1 k_3 \mathcal{N}}{\lambda \mathcal{D}} (T_1 + T_2).$$

Thus, for the non-linear model and general (Poisson) noise we have the total conductance:

$$\begin{aligned} \kappa_{0,1} = & \frac{k_1^2 \gamma}{2 (m k_1^2 + \gamma^2 (k + k_1))} + \\ & + \frac{3}{2} \gamma k_1 k_3 \frac{(2k + k_1) \gamma^2 \bar{T}}{(k + 2k_1) [\gamma^2 (k + k_1) + m k_1^2] + 2}. \\ & + 54 \frac{\gamma^4 k_1 k_3 \mathcal{N}}{\lambda \mathcal{D}} (T_1 + T_2). \end{aligned}$$

Observe that as the Gaussian limit is reached ($\lambda \rightarrow \infty$), the last contribution above vanish.

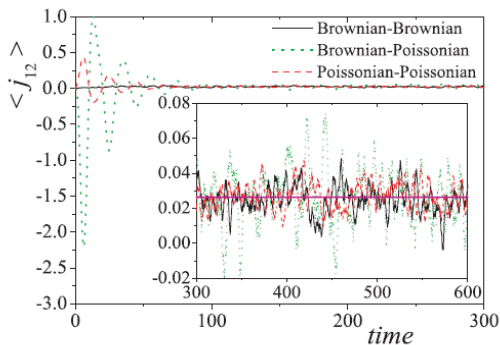


FIG. 1. (Color online) Average exchange flux $\langle j_{12} \rangle$ of a two-massive-particle system for different combinations of paradigmatic types of noise with $T_1 = 10$, $T_2 = 121/10$, $m = 10$, $\gamma = k = 1$, $k_1 = 1/5$, $k_3 = 0$, and $\lambda = 10$ for Poissonian particles. After the transient, κ agrees with the theoretical value, $\kappa = 21/800 = 0.02625$, with the fitting curves lying within the line thickness. The averages have been obtained by averaging over $850 \times (5 \times 10^5)$ points. The discretization used is $\delta t = 10^{-5}$ with snapshots at every $\Delta t = 10^{-3}$.

Thermal Conductance

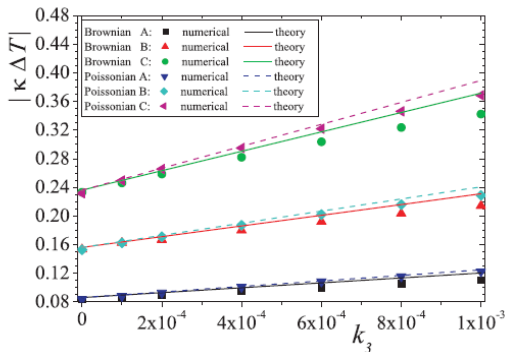


FIG. 2. (Color online) Comparison between numerically obtained values (symbols) and the first-order approximation of thermal conductance from Eqs. (8)–(10) for different temperature pairs, namely, $A = \{10, \frac{169}{10}\}$, $B = \{10, \frac{225}{10}\}$, and $C = \{10, \frac{289}{10}\}$ with $m = 10$, $\gamma = k = 1$, $k_1 = 1/5$, and $\lambda = 1$ for Poissonian particles.

Conclusions & perspectives

- We are working to systematize the case of a general potential.
- The method can be applied to a linear chain of arbitrary size, or a solid.
- Quantum version of the method.

Agradecimentos

