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▶ To cite this version:

Fabio Staniscia, Lev Truskinovsky. Passive viscoelastic response of striated muscles. 2021. hal-03100743

HAL Id: hal-03100743 https://hal.science/hal-03100743

Preprint submitted on 6 Jan 2021

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Passive viscoelastic response of striated muscles

Fabio Staniscia^{1,*} and Lev Truskinovsky²

¹LMS, École polytechnique, 91128 Palaiseau Cedex, France

²PMMH, CNRS - UMR 7636 PSL-ESPCI, 10 Rue Vauquelin, 75005 Paris, France

(Dated: June 21, 2019)

Muscle cells with sarcomeric structure exhibit highly nontrivial passive mechanical response. The difficulty of its continuum modeling is due to the presence of long-range interactions transmitted by extended protein skeleton. To build a rheological model for muscle 'material' we use a stochastic micromodel and derive a linear response theory for a half-sarcomere. Instead of the first order rheological equation, anticipated by A.V. Hill on the phenomenological grounds, we obtain a novel second order equation. We use the values of the microscopic parameters for frog muscles to show that the proposed rheological model is in excellent quantitative agreement with physiological experiments.

PACS numbers: 87.19.Ff, 46.35.+z, 87.15.A-, 87.85.jc

One of the simplest biological systems, that still defies the attempts to reproduce it artificially as a macroscopic material, is a striated muscle [1]. Its mechanical complexity is due to the presence of a large number of nonlinear, hierarchically organized microscopic subsystems that are strongly coupled through long-range interactions [2]. This makes the task of reconstructing the macroscopic constitutive relations describing even its passive mechanical response rather challenging [3, 4].

A broadly used phenomenological theory of the passive visco-elastic response of striated muscles, proposed by A.V.Hill [5–8], does not rely on coarse graining techniques [9, 10] and therefore does not offer a link between macro and micro parameters. Since Hill's rheological relation involves a single characteristic time scale, it also does not capture the difference in the passive response exhibited by striated muscles abruptly loaded in soft (isotonic) and hard (isometric) loading devices [11, 12].

A microscopically guided stochastic approach to muscle visco-elasticity was proposed by Huxley and Simmons [11, 13] who assumed that the individual force producing units (myosin cross-bridges) are stochastically independent. A mean-field interaction between the crossbridges was incorporated in a closely related model by Shimizu [14–17]. The two approaches have been recently unified [2, 18–20]. In the present article we use this framework to rigorously derive from a micro-model a linear rheological response theory for a muscle half-sarcomere. Our analysis builds on the work of Shiino [21] who obtained a similar linear response theory for the related model of Desai and Zwanzig [16, 17, 22]; other relevant out-of-equilibrium systems were studied in [23, 24].

Our main result is the linear spring-dashpot scheme which reproduces adequately the mechanical behavior of a muscle fiber subjected to a time dependent perturbation. In contrast to the classical model of Hill [5, 6], the proposed rheological equation contains not only the *first* but also the *second* time derivatives of the macroscopic displacement. We use the values of the microscopic parameters for frog muscles to show that our macroscopic model, which does not rely on any fitting parameters, is in excellent quantitative agreement with physiological experiment.



FIG. 1: Schematic representation of the transient response of a skeletal muscle subjected to an abrupt (a) isotonic and (b) isometric perturbation. Here σ is the stress and z is the elongation. The meaning of phases 1-4 is explained in the text, and the focus of this article is on phases 1 and 2. Adapted from [25].

The model contains two time scales that can be associated with the transient stages of muscle response known as phases 1 and 2, see Fig. 1 and [11, 26–28], with both of them captured adequately by our rheological model. The presence of the two time scales reflects involvement of the two parallel *passive* processes: the (microscopic) conformational change in N myosin heads and the macroscopic relaxation of the myofilaments in a viscous environment. The obtained linear rheological equation does not capture the collective barrier crossing which can be associated with the slower phase 3 explained in [2, 29, 30]; it also does not address the *active* phase 4, see Fig. 1.

We recall that striated muscle is a hierarchical chemomechanical system with the smallest scale represented by force generating half-sarcomeres [2, 31]. To reproduce the passive response of this molecular machine we use the simplest microscopic model developed in [18–20, 29, 30]. We assume that behind the muscle power stroke, responsible for fast force recovery, is a double well potential of Landau type, V(x), describing two conformational states of a cross-bridge [18]; it is known that myosin heads can be in other configurations [25, 27, 32], so here we simplify the real physical picture [12, 33]. We also assume that the potential V(x) is asymmetric, which allows the system to generate (stall) force in the physiological regime of isometric contractions. While this asymmetry is maintained actively [34], we can still interpret the short time response in such system as passive. The active response involving detachment of the cross-bridges becomes dominant at time-scales of the order of 40 ms [35].

To define the dimensionless units we rescale the length by the size of the maximum working stroke a, the energy by κa^2 , where κ is the cross-bridge stiffness, and the time by the ratio of the cross-bridge drag coefficient γ_x and κ . To model a bundle of thick and thin filaments linked by N cross-bridges and loaded with the force f we use the dimensionless energy [2]

$$H = \sum_{i=1}^{N} \left[V(x_i) + \frac{(y-x_i)^2}{2} + \frac{\lambda_f}{2} \left(z-y\right)^2 \right] - z f, \quad (1)$$

where the variables x_i represent the configuration of individual cross bridges. The latter are elastically coupled through the cross-bridge stiffness to a collective variable y representing an actin filament; the associated quadratic term in (1) represents mean field type interaction. The combined elasticity of actin and myosin filaments is described by the quadratic term coupling the variable y and z with the dimensionless coefficient λ_f representing the overall filamental stiffness, see [2] for more details.

We assume that the meso-scopic collective variable y(t) relaxes instantaneously [36, 37] and we can therefore eliminate such (Weiss-type) variable adiabatically. We obtain $y = (\langle x \rangle + \lambda_f z)/(1 + \lambda_f)$, where we introduced notation $\langle x \rangle = N^{-1} \sum_{i=1}^{N} x_i$. Substituting the obtained expression for y into (1) we obtain the redressed energy $\tilde{H}(\mathbf{x}, z) = H(\mathbf{x}, y(\mathbf{x}, z), z)$ whose main feature visà-vis the model of Huxley and Simmons is the presence of all-to-all interaction between the cross-bridges expressed through the term $\langle x \rangle^2$, see also [21, 22].

Next we assume that the macro-variable z(t) is also deterministic, however now its dynamics is governed by the relaxation equation: $\nu \dot{z} = -\partial \tilde{H}/\partial z$, where the superimposed dot denotes time derivative and ν is the dimensionless macroscopic friction coefficient obtained by normalizing the filamental drag coefficient γ_z by the myosin head drag coefficient γ_x . We can rewrite this equation as

$$\frac{\nu}{N}\dot{z} = \lambda_f \frac{\langle x \rangle - z}{1 + \lambda_f} + \frac{f}{N}.$$
(2)

The time scale $\tau_z = \nu (1 + \lambda_f)/(\lambda_f N)$ will then characterize the evolution of the macro-variable z.

In the evolution of the micro-variables x_i we need to account for the thermal noise: $\dot{x}_i = -\partial \tilde{H} / \partial x_i + \xi_i$, where

 $\langle \xi_i \rangle = 0$ and $\langle \xi_i(t) \xi_j(t') \rangle = (2/\beta) \,\delta(t - t') \,\delta_{ij}$. Here we introduced the inverse dimensionless temperature β . Under the assumption that N is large, the single particle probability density p(x,t) can be found from the nonlinear Fokker-Planck equation [21, 22, 38]

$$\frac{\partial p}{\partial t} = \frac{\partial}{\partial x} \left[\left(\frac{\partial V}{\partial x} + x - \frac{\langle x \rangle + \lambda_f z}{1 + \lambda_f} + \frac{1}{\beta} \frac{\partial}{\partial x} \right) p \right], \quad (3)$$

where we use approximation $\langle x \rangle \approx \int dx \, p \, x$. The stationary solution of (3) is $p_s(x) = Z^{-1} \exp \left[-\beta U\right]$, where

$$U(x,z) = V(x) + \frac{1}{2} \left(x - \frac{\langle x \rangle_s(z) + \lambda_f z}{1 + \lambda_f} \right)^2 \qquad (4)$$

is an effective double well potential and Z is a normalization constant [2]. In (4) we must use the self-consistence condition $\langle x \rangle_s = \int dx \, p_s(x) \, x$; such closure can be justified rigorously in the thermodynamic limit [39].

To develop the linear response theory we use as a starting point the approach developed in [21]. First, we linearize (3) to obtain the propagator

$$L p = \frac{\partial}{\partial x} \left[\left(\frac{\partial V}{\partial x} + x - \frac{\langle x \rangle_s + z \lambda_f}{1 + \lambda_f} + \frac{1}{\beta} \frac{\partial}{\partial x} \right) p \right].$$
(5)

A perturbation $\delta p(x,t) = p(x,t) - p_s(x)$ associated with a small change of the macroscopic strain variable $\delta z(t)$ will then satisfy a linear equation

$$\frac{\partial \delta p}{\partial t} = L \,\delta p - \frac{1}{1 + \lambda_f} \frac{\partial p_s}{\partial x} \left(\int \mathrm{d}x \, x \,\delta p + \lambda_f \,\delta z \right). \tag{6}$$

In particular, the collective variable $\langle \delta x \rangle(t) = \int dx \, x \, \delta p(x,t)$, will evolve according to

$$\langle \delta x \rangle(t) = \int_{-\infty}^{\infty} \frac{\langle \delta x(t') \rangle + \lambda_f \delta z(t')}{1 + \lambda_f} \,\chi_{xx}(t - t') \,\mathrm{d}t', \quad (7)$$

which in Fourier space reads $\langle \delta x \rangle(\omega) = \lambda_f \chi_{xx}(\omega) \delta z(\omega) / (1 + \lambda_f - \chi_{xx}(\omega))$. The susceptibility in (7) is defined by the relation

$$\chi_{xx}(t) = -\Theta(t) \int \mathrm{d}x \, x \, e^{L \, t} \frac{\partial}{\partial x} p_s(x) \,, \tag{8}$$

where $\Theta(t)$ is the Heaviside function.

Next we write the conventional fluctuation-dissipation type identity $\chi_{xx} = -\beta \Theta(t) dS_{xx}/dt$, where $S_{xx}(t) = \langle x(t) x(0) \rangle - \langle x \rangle_s^2$ is the auto-correlation function of a single element x_i which is assumed to be evolving in the effective potential (4). Given that in linear approximation each cross-bridge can be viewed as conducting independently a simple Brownian motion in a double well potential we can use the Kramers approximation to obtain [40, 41]:

$$S_{xx}(t) \simeq c \, e^{-\frac{t}{\tau_x}} \,, \tag{9}$$

where $\tau_x \simeq \frac{2\pi}{\sqrt{|U''(x_M)|}} \frac{e^{\beta U(x_M)}}{\sqrt{U''(x_0)} e^{\beta U(x_0)} + \sqrt{U''(x_1)} e^{\beta U(x_1)}}$, with $x_{0,1}$, being the minima of the potential (4) and x_M , the local maximum between them. The pre-factor can be also computed analytically $c = (x_0 - x_1)^2 q/(1+q)^2$ with $q = \sqrt{(U''(x_0)/U''(x_1))} \exp[\beta(U(x_0) - U(x_1))]$. In this computation we neglected the effect of the relaxation within a single well because it is much faster in physiological conditions than the barrier crossing.



FIG. 2: Schematic representation of the rheological model (11).

If we rewrite (9) in the Fourier space $\chi_{xx}(\omega) = c \beta/(1 + i \omega \tau_x)$, and use (2), we obtain the desired linear response relation between the macro-variables $\delta z(t)$ and $\delta f(t)$:

$$\begin{bmatrix} -\omega^2 \tau_x \, \tau_z + i \, \omega \left(\tau_z + \tau_x - \frac{\beta \, c \, \tau_z}{1 + \lambda_f} \right) + 1 - \beta \, c \end{bmatrix} \delta z(\omega) = \begin{bmatrix} i \, \omega \, \tau_x \, \frac{1 + \lambda_f}{\lambda_f \, N} + \frac{1 + \lambda_f - \beta \, c}{\lambda_f \, N} \end{bmatrix} \delta f(\omega) \,.$$
(10)

The corresponding spring-dash-pot model is shown in Fig. 2. At the inner level we have a parallel bundle of a spring with stiffness $E_1 = N (1 - \beta c)/(\beta c)$ and a dash-pot with viscosity $\eta_1 = N \tau_x/(\beta c)$. This subsystem accounts for the cross-bridge dynamics. The outer level contains a spring with stiffness $E = N \lambda_f/(1 + \lambda_f)$ and the dash-pot with viscosity $\eta = \nu$; the latter represents viscous response of the effective backbone.

In the real space the rheological relation (10) takes the form

$$\theta \,\nu \,\ddot{\delta z} + (\theta \,E + \nu) \,\dot{\delta z} + C \,\delta z = \delta f + \theta \,\dot{\delta f} \qquad (11)$$

where $C = E_1 E/(E_1+E)$ and $\theta = \eta_1/(E_1+E)$. If $\tau_x = 0$ we obtain the Kelvin-Voigt model, and if $\tau_z = 0$ Eq. 11 reproduces the rheological structure of the (passive) Hill's model.

To make a more detailed comparison in a hard device we can define the storage modulus G' and the loss modulus G'' as real and imaginary parts of the ratio $\delta f(\omega)/\delta z(\omega)$; in a soft device we must consider instead the ratio $\delta z(\omega)/\delta f(\omega)$. The frequency dependence of these parameters is illustrated in Fig. 3. Note the divergence of G''_H at large ω in qualitative difference with the Hill's model where this parameter tends to zero. Similarly, in the Hill's model G'_S has a finite limit at large ω while in our model it tends to zero.



FIG. 3: Frequency dependence of the loss and storage moduli in the hard and soft devices for the physiological choice of parameters.

Consider now the response of the system (11) to canonical step-like perturbations [42] as in the typical muscle experiments. In the hard device, the response to the input $\delta z(t) = \delta z_0 \Theta(t)$, is described by the equation $\delta f(t) =$ $\delta z_0 \left[(E-C) \ e^{-\frac{t}{\theta}} + C + \nu \ \delta(t) \right]$, were $\delta(t)$ is the Dirac delta function, whose effect cannot be detected in experiments unless the perturbation is strictly instantaneous. Note the first jump in tension $\lim_{t\to 0} \delta f(t) = E \,\delta z_0$, taking place simultaneously with the applied length step (phase 1 in Fig. 1b and 4a), which is the signature of a purely elastic behavior [12]. The elastic phase is followed by an exponential relaxation (phase 2 in Fig. 1b and 4a) with the time scale θ . The condition $\theta > 0$, or equivalently, $1 + \lambda_f - \beta c > 0$, then serves as a bound (since we are using an approximate expression for the correlation function S_{xx} for the mechanical stability threshold of the equilibrium system in the hard device.



FIG. 4: Response to a step like perturbation of the model (11) (analytical), compared with direct numerical simulations for N = 32768 crossbridges averaged over 10 realizations (numerics). Here f is the force per thick filament. (a): hard-device, (b): soft device.

In the soft-device the response to a small step like perturbation $\delta f(t) = \delta f_0 \Theta(t)$ is described by the equation

$$\delta z(t) = \delta f_0 \left[-\frac{\tau_+ \tau_-}{\nu (\tau_+ - \tau_-)} \left(1 - \frac{\tau_+}{\theta} \right) e^{-\frac{t}{\tau_+}} + \frac{\tau_+ \tau_-}{\nu (\tau_+ - \tau_-)} \left(1 - \frac{\tau_-}{\theta} \right) e^{-\frac{t}{\tau_-}} + \frac{1}{C} \right], \quad (12)$$

where we introduced two new effective time scales

$$\tau_{\pm} = \frac{\eta_1 + \nu}{2E_1} + \frac{\nu}{2E} \pm \sqrt{\left(\frac{\eta_1 + \nu}{2E_1} - \frac{\nu}{2E}\right)^2 + \frac{\nu^2}{E_1E}}.$$
 (13)

Observe that according to (12), $\lim_{t\to 0} \delta z(t) = 0$, indicating that there is no synchronous response to a steplike perturbation. The fact that in the soft device the relaxation starts when the perturbation is still being delivered [27], explains the difficulty in separating stages 1 and 2 (shown in Fig. 1a and 4b) of the transient response in isotonic conditions: since $\tau_- \ll \tau_+$ the first relaxation process with the time scale τ_- was sometimes interpreted as purely elastic [11, 26–28]. The approximate stability condition is now $1 - \beta c > 0$. Note the difference between the stability thresholds in soft and hard device reflecting the ensemble inequivalence in this mean field system [20, 43].

Using the basic physiological constraints on the parameters, $\beta \ge 0$, $\nu \ge 0$, $\lambda_f \ge 0$, one can show that in stable regimes $\tau_+ > \theta$, which is in agreement with the fact that the relaxation in the soft the device is slower than in the hard device. In unstable regimes the expression under the square root in (13) may become negative which would indicate the possibility of oscillatory relaxation (in the soft device). Damped oscillations have been indeed observed in some mechanical experiments [44, 45], however, they appeared at larger timescales where the implied mechanical instability might have been suppressed actively [34].

To make quantitative predictions and compare our results with physiological measurements we need to calibrate the model using the physical values of parameters.

For the cross-bridge stiffness we take the value $\kappa = 3.29 \text{ pN/nm}$ [46], while the combined stiffness of actin and myosin filaments can be estimated at the value $\kappa_f \simeq 153 \text{ pN/nm}$ [36, 37]. Since $N \simeq 100$ we obtain $\lambda_f = \kappa_f/(\kappa N) \simeq 0.46$. Given that the maximum working stroke is a = 11 nm, the energy scale is $\kappa a^2 = 398 \text{ zJ}$, and we can conclude that at T = 273 K the value of the dimensionless inverse temperature is $\beta \simeq 105$.

To estimate parameter γ_x we assume that a typical myosin head has the diameter $h \simeq 6$ nm and that the cytoplasm has the effective dynamical viscosity $\eta \simeq$ 2.3×10^{-3} Pas [47, 48]. Viewing it as a sphere we obtain $\gamma_x \simeq 3 \pi \eta h \simeq 1, 30 \times 10^{-4}$ ms pN/nm, so that the corresponding time unit is $\gamma_x/\kappa = 3.44 \times 10^{-5}$ ms. The dimensionless viscosity of a thick filament $\nu = \gamma_z/\gamma_x$ can be now calculated using the estimate for the drag coefficient $\gamma_z = \xi/\rho_f \simeq 0.480 \,\mathrm{ms} \,\mathrm{pN/nm}$, where $\xi \simeq 3 \times 10^8 \,\mathrm{N} \,\mathrm{s/m^3}$ is a viscosity coefficient obtained in [46] for *Rana temporaria*. We use here the expression $\rho_f = 2/(\sqrt{3} \, d^2)$ for the density of thick filaments in the section of a sarcomere where the thick filaments form a triangular pattern and lie at a distance $d = 43 \,\mathrm{nm}$ [49]. The drag coefficient for the bundle of thick filaments constituting the half-sarcomere can be taken in the form $\gamma_z^{HS} \simeq \xi S$, where $S \simeq 8 \,\mu\mathrm{m}^2$ is the cross section of the sarcomere. Substituting this expression into the estimate for γ_z we can extend the predictions of the model to the case of the whole half-sarcomere. The number of crossbridges N has to be then multiplied by number of thick filaments in a half-sarcomere with the parallel rescaling $\delta f \to \delta f/(S \,\rho_f)$.

To model the double well potential V(x) we use the simplest quartic function with one minimum V(0) = 0 (pre-power-stroke state) and another minimum V(-0.25) = -0.03 (post-power-stroke state). The barrier is chosen at the level $V(x_*) = 0.11$ because the activation energy for the muscle power stroke was previously estimated to be either 55.3 zJ or 95.1 zJ [50, 51] and we have chosen the smaller value as more relevant for the the fast time transient response.

The rheological equation (11) obtained for a single halfsarcomere must be now renormalized to the scale of a muscle fiber. To this end we assume that the response is affine, at least when perturbations are sufficiently small. We can view a myofibril as a chain of $L \sim 10^4$ halfsarcomeres connected in series, and represent a muscle fiber by a parallel arrangement of $M \sim 200 - 2000$ such myofibrils. The renormalization will then reduce to the substitutions $\delta z \to \delta z/L$ and $\delta f \to \delta f/M$.

Using these values of parameters we compute $\theta \simeq 10200 \ (\simeq 0.35 \text{ ms})$ which is compatible with the relaxation time measured in [26]. Next we estimate the elastic modulus for the instantaneous response of a half-sarcomere in the hard device. Given that $l_0 = 95.5 \ (\simeq 1.05 \,\mu\text{m})$ is the length of a half-sarcomere and $r \simeq 0.83$ is the ratio of the cross-section of the muscle fiber occupied by sarcomeres [52], we obtain $E_Y = \lim_{t\to 0} \delta f(t) \rho_f r l_0 / \delta z_0 \simeq N \rho_f r \lambda_f l_0 / (1 + \lambda_f) = 190$ ($\simeq 57 \text{ MPa}$), close to the value measured in [26]. Finally we compute $\tau_+ \simeq 12700 \ (\simeq 0.44 \text{ ms})$ and $\tau_- \simeq 116 \ (\simeq 0.004 \text{ ms})$, which is also in good agreement with experimental observations [27, 53].

We are now in the position to evaluate to what extent the microscopic stochastic model and the macroscopic deterministic model can reproduce the outcomes of the realistic experiments. Numerical simulations of the microscopic model were conducted with a second order stochastic Runge-Kutta algorithm. We simulated the response of the microscopic model to a step-like perturbation δz in a hard device, and δf in a soft device, computing the corresponding responses $\delta f(t)$ and $\delta z(t)$.

The results are summarized in Fig. 4 where we used the physiological values of parameters except that in the simulation of the stochastic micro-model of a half sarcomere, instead of the realistic value $N \sim 500000$, we used the computationally reachable value N = 32768, while appropriately rescaling the parameter ν to ensure that the time scale τ_z remains at its realistic value. Numerical experiments aimed at a single bundle of thick and thin filaments with N = 128 and using the physiological value of ν give practically the same results. In both cases the agreement between the stochastic model and the rheological equation (11) is excellent for both hard and soft devices.

To conclude, we have shown that the time dependent passive viscoelastic behavior of striated muscles can be understood at the quantitative level starting from the microscopic structure of a half sarcomere. From a stochastic microscale model we derived a deterministic rheological model which describes linear response of muscle fibers under various loading conditions and found explicit relations between the macroscopic and the microscopic parameters. The fact that the derived rheological relation involves not only first but also second derivatives allowed us to explain the qualitative differences in the mechanical response of isometrically and isotonically loaded muscles. The model was calibrated based on independent data and excellent quantitative agreement was reached with physiological observations. It would be of interest to check experimentally the predicted link between the mechanical response to fast perturbations [26, 53] and the power spectrum of mechanical fluctuations [54].

Acknowledgements. The authors thank M. Caruel, R. Garcia Garcia and S. Ruffo for helpful discussions. FS was supported by a postdoctoral fellowship from Ecole Polytechnique. The work of LT was supported by the grant ANR-10-IDEX-0001-02 PSL.

* fabiostaniscia@gmail.com

- N. Kidambi, R. L. Harne, and K.-W. Wang, Phys. Rev. E 98, 043001 (2018).
- [2] M. Caruel and L. Truskinovsky, Reports on Progress in Physics 81, 036602 (2018).
- [3] M. Caruel, P. Moireau, and D. Chapelle, Biomechanics and Modeling in Mechanobiology 18, 563 (2019).
- [4] F. Regazzoni, L. Dedè, and A. Quarteroni, Biomechanics and Modeling in Mechanobiology 17, 1663 (2018).
- [5] A. V. Hill, Proceedings of the Royal Society of London B: Biological Sciences 126, 136 (1938).
- [6] A. V. Hill, Proceedings of the Royal Society of London. Series B-Biological Sciences 136, 399 (1949).
- [7] S. A. Glantz, Journal of Biomechanics 7, 137 (1974).
- [8] B. Gerazov and P. N. Garner, in Speech and Computer, edited by A. Ronzhin, R. Potapova, and G. Németh (Springer International Publishing, Cham, 2016) pp. 84– 91.
- [9] G. P. DeVault and J. A. McLennan, Phys. Rev. 137, A724 (1965).
- [10] F. Bavaud, Journal of Statistical Physics 46, 753 (1987).

- [12] M. Reconditi, M. Linari, L. Lucii, A. Stewart, Y.-B. Sun, P. Boesecke, T. Narayanan, R. F. Fischetti, T. Irving, G. Piazzesi, M. Irving, and V. Lombardi, Nature 428, 578 EP (2004).
- [13] E. Eisenberg, T. Hill, and Y. Chen, Biophysical Journal 29, 195 (1980).
- [14] H. Shimizu and T. Yamada, Progress of Theoretical Physics 47, 350 (1972).
- [15] K. Kometani and H. Shimizu, Journal of Statistical Physics 13, 473 (1975).
- [16] L. L. Bonilla, Journal of Statistical Physics 46, 659 (1987).
- [17] T. Frank, Physics Letters A **329**, 475 (2004).
- [18] L. Marcucci and L. Truskinovsky, Phys. Rev. E 81, 051915 (2010).
- [19] L. Marcucci and L. Truskinovsky, The European Physical Journal E 32, 411 (2010).
- [20] M. Caruel, J.-M. Allain, and L. Truskinovsky, Phys. Rev. Lett. **110**, 248103 (2013).
- [21] M. Shiino, Phys. Rev. A 36, 2393 (1987).
- [22] R. C. Desai and R. Zwanzig, Journal of Statistical Physics 19, 1 (1978).
- [23] A. Patelli, S. Gupta, C. Nardini, and S. Ruffo, Phys. Rev. E 85, 021133 (2012).
- [24] A. Patelli and S. Ruffo, The European Physical Journal D 68, 329 (2014).
- [25] A. F. Huxley, The Journal of Physiology 243, 1 (1974).
- [26] G. Piazzesi, M. Reconditi, N. Koubassova, V. Decostre, M. Linari, L. Lucii, and V. Lombardi, The Journal of Physiology 549, 93 (2003).
- [27] V. Decostre, P. Bianco, V. Lombardi, and G. Piazzesi, Proceedings of the National Academy of Sciences 102, 13927 (2005).
- [28] Cell **131**, 784 (2007).
- [29] M. Caruel and L. Truskinovsky, Phys. Rev. E 93, 062407 (2016).
- [30] M. Caruel and L. Truskinovsky, Journal of the Mechanics and Physics of Solids 109, 117 (2017).
- [31] B. Alberts, D. Bray, K. Hopkin, A. D. Johnson, J. Lewis, M. Raff, K. Roberts, and P. Walter, *Essential cell biology* (Garland Science, 2013).
- [32] M. Caremani, L. Melli, M. Dolfi, V. Lombardi, and M. Linari, The Journal of Physiology **591**, 5187 (2013).
- [33] J. E. Molloy, J. E. Burns, J. Kendrick-Jones, R. T. Tregear, and D. C. S. White, Nature **378**, 209 (1995).
- [34] R. Sheshka, P. Recho, and L. Truskinovsky, Phys. Rev. E 93, 052604 (2016).
- [35] J. Howard, Mechanics of Motor Proteins and the Cytoskeleton (Sinauer Associates, Publishers, 2001).
- [36] H. Huxley, A. Stewart, H. Sosa, and T. Irving, Biophysical Journal 67, 2411 (1994).
- [37] K. Wakabayashi, Y. Sugimoto, H. Tanaka, Y. Ueno, Y. Takezawa, and Y. Amemiya, Biophysical Journal 67, 2422 (1994).
- [38] T. D. Frank, Nonlinear Fokker-Planck equations: fundamentals and applications (Springer Science & Business Media, 2005).
- [39] T. Dauxois, S. Lepri, and S. Ruffo, Communications in Nonlinear Science and Numerical Simulation 8, 375 (2003), chaotic transport and compexity in classical and quantum dynamics.
- [40] J. L. Skinner and P. G. Wolynes, The Journal of Chemical

Physics 69, 2143 (1978).

- [41] R. Pratolongo, A. Perico, K. F. Freed, and A. Szabo, The Journal of Chemical Physics 102, 4683 (1995).
- [42] F. Mainardi and G. Spada, The European Physical Journal Special Topics 193, 133 (2011).
- [43] J. Barré, D. Mukamel, and S. Ruffo, Phys. Rev. Lett. 87, 030601 (2001).
- [44] K. A. P. Edman and N. A. Curtin, The Journal of Physiology 534, 553 (2001).
- [45] H. Sugi and T. Tsuchiya, J Physiol **319**, 219 (1981), 7320912[pmid].
- [46] L. E. Ford, A. F. Huxley, and R. M. Simmons, The Journal of Physiology 269, 441 (1977).
- [47] M. J. Kushmerick and R. J. Podolsky, Science 166, 1297 (1969).
- [48] M. Arrio-Dupont, S. Cribier, G. Foucault, P. Devaux,

and A. d'Albis, Biophysical Journal 70, 2327 (1996).

- [49] I. Matsubara and G. F. Elliott, Journal of Molecular Biology 72, 657 (1972).
- [50] R. Elangovan, M. Capitanio, L. Melli, F. S. Pavone, V. Lombardi, and G. Piazzesi, The Journal of Physiology 590, 1227 (2012).
- [51] M. Anson, Journal of Molecular Biology 224, 1029 (1992).
- [52] B. A. Mobley and B. R. Eisenberg, The Journal of General Physiology 66, 31 (1975).
- [53] G. Piazzesi, L. Lucii, and V. Lombardi, The Journal of Physiology 545, 145 (2002).
- [54] E. Malmerberg, C. A. Kerfeld, and P. H. Zwart, IUCrJ 2, 309 (2015).