# Transition to superdiffusive behavior in intracellular actin-based transport mediated by molecular motors

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Intracellular transport of large cargoes, such as organelles, vesicles, or large proteins, is a complex dynamical process that involves the interplay of adenosine triphosphate–consuming molecular motors, cytoskeleton filaments, and the viscoelastic cytoplasm. In this work we investigate the motion of pigment organelles (melanosomes) driven by myosin-V motors in *Xenopus laevis* melanocytes using a high-spatio-temporal resolution tracking technique. By analyzing the obtained trajectories, we show that the melanosomes mean-square displacement undergoes a transition from a subdiffusive to a superdiffusive behavior. A stochastic theoretical model, which explicitly considers the collective action of the molecular motors, is introduced to generalize the interpretation of our data. Starting from a generalized Langevin equation, we derive an analytical expression for the mean square displacement, which also takes into account the experimental noise. By fitting theoretical expressions to experimental data we were able to discriminate the exponents that characterize the passive and active contributions to the dynamics and to estimate the "global" motor forces correctly. Then, our model gives a quantitative description of active transport in living cells with a reduced number of parameters.

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## I. INTRODUCTION

Molecular motors are responsible for the active transport of organelles and other cargoes along cytoskeleton tracks to their correct destination in the cytoplasm. Three different classes of molecular motors are involved in this task: dynein and kinesin motors, which transport cargoes toward the minus and plus ends of microtubules, respectively, and myosin motors, responsible for the transport along actin filaments toward the barbed end (reviewed in [1]). While properties of microtubule-dependent transport, considered being responsible for long-distance transport, have been extensively studied *in vivo* [2–5], we still do not know the important aspects of the properties and regulation of actin-dependent transport which are believed to support local, short-distance movement of cargoes in living cells [6].

The cytoskeleton is a crowded network of semiflexible linear protein polymers with a complex dynamics, that can exert forces and affect the rheology of cells [7,8]. Particularly, *F*-actin network has no global directionality and consists on intercepting randomly distributed filaments [9]. The cytoskeleton, organelles, proteins, and motors are part of the highly crowded cytoplasm where the transport of cargoes takes place.

Active and passive microrheology techniques are two distinct approaches to study the mechanical properties of the intracellular environment [10]. While in active microrheology experiments the embedded probe particles are manipulated by external fields and their displacement is measured, the spontaneous displacement fluctuations of probe particles is analyzed in passive microrheology experiments.

Recent studies using both microrheology techniques have shown that the action of adenosine triphosphate (ATP) –consuming molecular motors drives the system out of equilibrium. The violation of the fluctuation-dissipation theorem has been observed in an *in vitro* system consisting of a crosslinked actin network with embedded myosin motors [7] and also for beads attached to the cell membrane of human airway smooth muscle cells [8]. Likewise, direct evidence of the deviation from equilibrium inside the cell has been recently obtained [11,12]. In Ref. [12], both forced and spontaneous motions of magnetic microbeads engulfed by *Dyctyostelium* cells are analyzed to derive the power spectrum of forces acting on intracellular phagosomes. However, the theoretical model proposed to describe their data did not distinguish between active and passive contribution to the transport.

In most of passive microrheology experiments it has been observed that the mean-square displacement (MSD) of the probes displays a transition from a subdiffusive to a super-diffusive regime [8,13–18]. The subdiffusive behavior is characterized by an exponent ranging from 0.2 [19] to  $\lesssim 1$  [20,21] while the superdiffusive behavior presents exponents close to 1.5 [8,12]. Typically, the transition time between these two regimes is on the order of 1 s.

While some authors attribute the subdiffusive behavior to elastic trapping [20], obstruction [22], crowding [23], or stalling [8], others propose that apparent subdiffusion can arise from the noise inherent to single particle tracking (SPT) experiments due to slight errors on the determination of the actual particle position [24]. Although there is not a general consensus of the causes of subdiffusion yet, it is well accepted that superdiffusion is originated in the collective behavior of molecular motors [8,11,12].

Dynamical properties of motor proteins have been well

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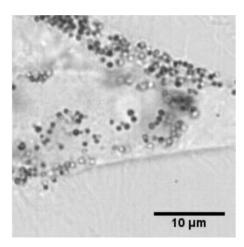


FIG. 1. A *Xenopus laevis* melanophore image obtained using bright-field transmission light microscopy. Melanosomes can be clearly identify and followed with high-spatio-temporal resolution using a SPT technique.

characterized *in vitro* but not in the viscous environment of a cell. Recently, pioneering works developed new techniques that allow to measure the forces exerted by molecular motors directly [25,26]. In Ref. [26], motion enhanced differential interference contrast movies of living NT2 (neuron-committed teratocarcinoma) cells was used to determine the force-velocity curves *in vivo*. In Ref. [25], the force exerted by kinesin-1 on lipid droplets in fly embryos was measured using a novel development of optical traps that can be used to apply precise forces to moving droplets *in vivo*. Unexpected differences between motor regulation *in vivo* and *in vitro* were found, revealing that further investigation should be done in the field.

Melanophore cells are an exceptionally convenient model system to study intracellular transport driven by molecular motors [27], and thus to investigate some properties of out-of-equilibrium systems in living cells. Melanocytes have pigment organelles called melanosomes which are filled with the black-pigment melanin. Then, they can be easily imaged using bright-field transmission-light microscopy (Fig. 1) and tracked with millisecond temporal resolution and nanometer precision [3]. Using a SPT technique [2,28], we follow the motion of myosin-V driven melanosomes along actin filaments and compute the MSD from the analysis of their trajectories.

Melanosomes are more or less spherical and stiff particles with sizes around 500 nm. Pigment organelles can be distributed in the cells in two configurations: either aggregated in the perinuclear region or homogeneously dispersed in the cytoplasm. The transport of pigment organelles during aggregation and dispersion is regulated by signaling mechanisms initiated by the binding of specific hormones to cell surface receptors [29]. In a recent paper, we analyzed the MSD dependence with the time lag using an empirical model [13]. With the aid of numerical simulations, we concluded that the main difference between aggregation and dispersion condition was the average time the melanosome spends diffusing during the intervals between successive processive motions. As a consequence, active part of the transport would not be influenced by stimulation conditions.

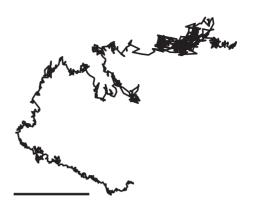


FIG. 2. A representative trajectory of a melanosome followed during 70 s with 0.07 s resolution. The cell was stimulated with 100 nM of MSH. Scale bar=0.5  $\mu$ m.

In this work we explore an alternative and complementary approach to obtain quantitative information about the intracellular transport that can be generalized to other biological data. Starting from a stochastic model that explicitly considers the viscoelastic properties of the intracellular environment and the action of molecular motors we derive an analytical expression for the MSD which also takes into account the experimental noise. By fitting the logarithmic derivative of the MSD-versus-lag time to the experimental data we were able to discriminate the passive and active contributions to the melanosome dynamics. Experiments using cells where melanosomes are not being actively transport by myosin-V confirmed the robustness of the model. Furthermore, our approach enables us to quantitatively assess the motor forces correlation function, to estimate the magnitude of motor forces and to determine an effective diffusion coefficient by only using data coming from a passive microrheology experiment.

## II. EXPERIMENTS

Melanophores were treated with nocodazole as described in the Appendix in order to depolymerize microtubules. After this treatment, aggregation and dispersion of melanosomes were induced by addition of melatonin and melanocytestimulating hormones (MSH), respectively. Movies of regions of the cells were recorded from which a total of 134 trajectories of melanosomes moving along actin filaments were obtained in aggregation and dispersion by using the pattern-recognition algorithm [3]. Figure 2 shows a representative trajectory of a myosin-V driven melanosome during dispersion.

The MSD for every trajectory was calculated as

$$\rho(\tau) = \langle \lceil x(t+\tau) - x(t) \rceil^2 + \lceil y(t+\tau) - y(t) \rceil^2 \rangle, \tag{1}$$

where x and y are the coordinates of the particle,  $\tau$  is the time lag and the brackets represents the time average.

Figure 3 shows the average value of the MSD obtained during dispersion and aggregation. The average distance traveled by the organelles during dispersion is significantly higher than during aggregation as was previously observed [2,9]. For times larger than 1 s, the behavior presents anomalous diffusion with an exponent around 1.35.

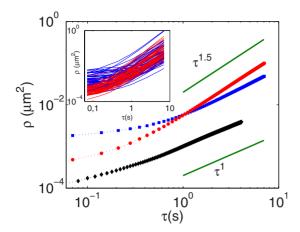


FIG. 3. (Color online) Double-logarithmic plot of the MSD as a function of the time lag. Experimental data points for wild-type cells are symbolized by squares (aggregation) and circles (dispersion), while the data corresponding to mutant cells are symbolized by rhomboids. The symbols represent an average over all the trajectories. Inset: MSD vs time lag plot for all trajectories in wild-type cells.

The observation of the residual value of the MSD at  $\tau \to 0$  ( $\rho_o$ ) suggests that transported melanosomes move in a constrained region with a dynamics faster than the temporal resolution of the tracking method. The presence of this jittering also results in an apparent subdiffusive behavior for short times scales [24,30].

We compute  $\rho_o$  by linear extrapolation of the first four data points of each experimental  $\rho$  vs  $\tau$  plot. The obtained  $\rho_o$  values followed an exponential distribution, with characteristic value equal to  $13\pm6\times10^{-4}~\mu\text{m}^2$  for aggregation and  $2.6\pm0.1\times10^{-4}~\mu\text{m}^2$  for dispersion.

It is important to mention that the MSD vs  $\tau$  plots for the different trajectories present a wide distribution, principally related to the spreading of the  $\rho_o$  values (see the inset in Fig. 3). This observation was also found in other systems [12,31]. In this sense, the ensemble average plot shown in Fig. 3 represents a poor estimator of the behavior of the driven particle. Instead, the logarithmic derivative of the MSD-versus-lag time  $\beta(\tau)$ , defined as [15]

$$\beta(\tau) = \frac{d}{d \ln \tau} \ln \rho(\tau), \qquad (2)$$

redounds on a more accurate magnitude to describe the dynamics of the transport. Figure 4 shows  $\beta$  as a function of the time lag for the different experimental conditions.

Furthermore, we investigate the behavior of melanosomes when the activity of myosin-V motors is inhibited. This was achieved using cells with mutated myosin-V which cannot attach to actin filaments (see Appendix). In these conditions, trajectories were recorded and the corresponding average behavior of the MSD is shown in Fig. 3. We also analyzed the value of the MSD obtained by extrapolation at  $\tau$ =0 s for the dominant-negative cells and verified that the  $\rho_o$  values obtained followed a Gaussian distribution with 0.56  $\times$  10<sup>-4</sup>  $\mu$ m<sup>2</sup> mean and  $\sigma$ =0.15 $\times$ 10<sup>-4</sup>  $\mu$ m<sup>2</sup>. These values

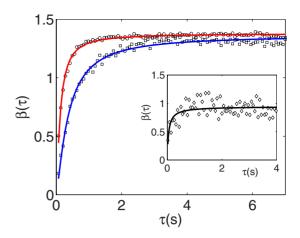


FIG. 4. (Color online) The logarithmic derivative of the MSD as a function of the time lag [Eq. (2)]. The experimental data for wild-type cells are symbolized with squares (aggregation) and circles (dispersion). The solid lines represent the fit to  $\beta$  of the expression given by Eq. (23). Inset: The same plot for mutant cells. The solid line represents the fit of Eq. (25). The symbols represent ensemble averages.

are in agreement with the ones expected according to the error of the tracking method [13,24].

## III. THEORETICAL MODEL

In this section we apply a quantitative stochastic model introduced in Ref. [32]. We consider the diffusive behavior of a melanosome of mass m, immersed in the intracellular medium and simultaneously driven by molecular motors. For this purpose, we describe the resulting dynamics by means of the generalized Langevin equation (GLE)

$$m\ddot{X}(t) + \int_0^t dt' \, \gamma(t - t') \dot{X}(t') = F(t), \tag{3}$$

where  $\gamma(t)$  is the dissipative memory kernel that characterizes the viscoelastic properties of the cytoplasm. The random force F(t) is assumed to be the sum of two uncorrelated contributions

$$F(t) = \xi(t) + \chi(t) \tag{4}$$

being  $\xi(t)$  the *internal* noise due to thermal activity, while  $\chi(t)$  is an *external* random force representing the action of the molecular motors.

The thermal noise  $\xi(t)$  is a zero centered and stationary random force with correlation function

$$\langle \xi(t)\xi(t')\rangle = C(|t-t'|),\tag{5}$$

and is related to the memory kernel  $\gamma(t)$  via the fluctuation-dissipation theorem [33]

$$C(t) = k_B T \gamma(t), \tag{6}$$

where T is the absolute temperature, and  $k_B$  is the Boltzmann constant. This internal noise is the responsible for the passive transportation. To reproduce the subdiffusive behavior due to the viscoelastic properties of the cytoplasm, the noise auto-

correlation function (5) can be modeled as a power law [34]

$$C(t) = \frac{C_0}{\Gamma(1-\lambda)} \left(\frac{t}{\tau_0}\right)^{-\lambda},\tag{7}$$

where  $0 < \lambda < 1$ ,  $C_0$  is a proportionality coefficient,  $\tau_0$  is set to 1 s, and  $\Gamma(z)$  is the Gamma function.

On the other hand, the random force  $\chi(t)$  is not related to the memory kernel  $\gamma(t)$  because it stems from the activity of the molecular motors. The irreversible conversion of chemical energy from ATP hydrolysis into the particle motion via the activity of myosin-V motors drives the system to an out-of-equilibrium situation. As a consequence, the fluctuation-dissipation theorem is no longer valid.

Assuming that the actin network has no global directionality, the random force  $\chi(t)$  exerted by the action of the myosin-V motor is chosen as a zero-centered one. Moreover, in recent experiments it was observed that the forces power spectrum exhibits a power-law behavior [8,11,12,14]. Then, we assume that the autocorrelation function

$$\langle \chi(t)\chi(t')\rangle = \Lambda(|t-t'|)$$
 (8)

can be written as [32]

$$\Lambda(t) = \frac{\Lambda_0}{\Gamma(1-\alpha)} \left(\frac{t}{\tau_0}\right)^{-\alpha},\tag{9}$$

where  $0 < \alpha < 1$ ,  $\Lambda_0$  is a proportionality coefficient and  $\tau_0$  is set to 1 s.

To evaluate the time scale of inertial effects, we consider that the average melanosome diameter is 500 nm [35], its density is  $\sim$ 1.2 g/ml [36] and  $\gamma \sim 10^{-5}$  Ns/m (see references in [37]). Then, the characteristic time is  $\sim 10^{-12}$  s and the inertial effects can be neglected. In this situation, the temporal evolution of the displacement can be obtained from (3), leading to

$$X(t) = x_0 + \int_0^t dt' G(t - t') [\xi(t') + \chi(t')], \qquad (10)$$

where  $x_0=X(t=0)$  is the deterministic initial position of the particle. The relaxation function G(t) is the inverse form of the Laplace transform

$$\hat{G}(s) = \frac{1}{s\,\hat{\gamma}(s)},\tag{11}$$

where  $\hat{\gamma}(s)$  is the Laplace transform of the dissipative memory kernel.

To obtain an analytical expression of the MSD it is necessary to calculate the two-time position correlation. Starting from Eq. (10) and using relation (6), we have

$$\begin{split} \langle X(t+\tau)X(t)\rangle &= x_0^2 + k_B T \big[ I(t) + I(t+\tau) - I(\tau) \big] \\ &+ \int_0^t dt_1 \big[ G(t_1)H(t_1+\tau) + G(t_1+\tau)H(t_1) \big], \end{split} \tag{12}$$

where the involved relaxation functions are given by

$$I(t) = \int_0^t dt' G(t'), \tag{13}$$

$$H(t) = \int_0^t dt' G(t') \Lambda(t - t'). \tag{14}$$

I(t) only depends on the internal noise, while H(t) is related to both internal and external contributions.

Using the autocorrelation functions (7) and (9), the involved relaxation functions can be written as

$$I(t) = \frac{k_B T}{C_0} \frac{1}{\Gamma(\lambda + 1)} \left(\frac{t}{\tau_0}\right)^{\lambda},\tag{15}$$

$$G(t) = \frac{k_B T}{\tau_0 C_0} \frac{1}{\Gamma(\lambda)} \left(\frac{t}{\tau_0}\right)^{\lambda - 1},\tag{16}$$

$$H(t) = \varepsilon k_B T \frac{1}{\Gamma(\lambda - \alpha + 1)} \left(\frac{t}{\tau_0}\right)^{\lambda - \alpha},\tag{17}$$

where

$$\varepsilon = \frac{\Lambda_0}{C_0} \tag{18}$$

is a dimensionless parameter that measures the relative intensity among the motors forces and the thermal forces.

For  $2\lambda - \alpha > 0$  the integral term of (12) can be explicitly evaluated. In this case, it can be demonstrated that the MSD long time limit

$$\rho(\tau) = \lim_{t \to \infty} \langle [X(t+\tau) - X(t)]^2 \rangle \tag{19}$$

can be written as [32]

$$\rho(\tau) = \frac{2k_B T}{\gamma_0} \left\{ \frac{1}{\Gamma(\lambda + 1)} \left( \frac{\tau}{\tau_0} \right)^{\lambda} + \varepsilon K_{\lambda, \alpha} \left( \frac{\tau}{\tau_0} \right)^{2\lambda - \alpha} \right\}, \quad (20)$$

where

$$K_{\lambda,\alpha} = \Gamma(\alpha - 2\lambda) \frac{\sin[\pi(\lambda - \alpha)] - \sin(\pi\lambda)}{\pi}, \qquad (21)$$

is a positive constant and  $\gamma_0 = C_0/kT$ . Note that Eq. (20) represents the MSD in the one-dimensional case, while the obtained trajectories are two-dimensional. However, the isotropic two-dimensional version of Eq. (20) can be readily obtained by replacing the factor  $2k_BT/\gamma_0$  by  $4k_BT/\gamma_0$ .

It is worth mentioning that, although  $0 < \alpha < 1$ , when  $1 < 2\lambda - \alpha < 2$  the MSD (20) exhibits a crossover from a sub-diffusive regime with an exponent  $\lambda$ , to a superdiffusive regime with an exponent  $2\lambda - \alpha$  [32]. The first term of (20) represents the passive transport of the melanosome in the viscoelastic medium, while the second one corresponds to the contribution of the random force  $\chi(t)$  originated in the activity of the myosin-V motors.

Finally, to compare the analytical expression (20) with the experimental data obtained using SPT techniques, it is necessary to take into account the error on the particle position determination. As established in Refs. [24,38], the effect of

TABLE I. Dimensionless parameters of the model  $\lambda$ ,  $\alpha$ ,  $\epsilon$ , and  $\delta$  were obtained from the fitting of Eqs. (23) and (25) to the experimental data shown in Fig. 4. The values are given as mean  $\pm$  standard deviation of the ensemble.

	λ	α	ε	δ
Aggregation	$0.96 \pm 0.04$	$0.58 \pm 0.08$	$202 \pm 38$	$41 \pm 8$
Dispersion	$0.98 \pm 0.03$	$0.58 \pm 0.10$	$83 \pm 13$	$4.0\pm0.6$
Mutant m-V	$0.94 \pm 0.04$			$0.061 \pm 0.014$

an uncorrelated noise of variance  $\eta^2$  generated by measurement errors in particle location or by biological activity is to add a constant to the MSD. Then,  $\rho(\tau)$  given in (20) rewrites as

$$\rho(\tau) \to \rho(\tau) + 2\,\eta^2. \tag{22}$$

As it was previously discussed, the corresponding local slope of the MSD-versus-lag time (2) is a better function to characterize the dynamics. From (20) and (22) we get

$$\beta(\tau) = \frac{\frac{\lambda}{\Gamma(\lambda+1)} \left(\frac{\tau}{\tau_0}\right)^{\lambda} + \varepsilon(2\lambda - \alpha) K_{\lambda,\alpha} \left(\frac{\tau}{\tau_0}\right)^{2\lambda - \alpha}}{\frac{1}{\Gamma(\lambda+1)} \left(\frac{\tau}{\tau_0}\right)^{\lambda} + \varepsilon K_{\lambda,\alpha} \left(\frac{\tau}{\tau_0}\right)^{2\lambda - \alpha} + \delta}, \tag{23}$$

where

$$\delta = \gamma_0 \frac{(2\,\eta)^2}{4k_B T}.\tag{24}$$

In particular, setting  $\varepsilon=0$  in Eq. (23) one gets

$$\beta(\tau) = \frac{\lambda \left(\frac{\tau}{\tau_0}\right)^{\lambda}}{\left(\frac{\tau}{\tau_0}\right)^{\lambda} + \Gamma(\lambda + 1)\delta},\tag{25}$$

which corresponds to the absence of active transport case.

## IV. RESULTS AND DISCUSSION

In this paper we study the motion of 500 nm melanosomes in *Xenopus laevis* melanophores treated with nocodazole (a microtubule depolymerizer), where the active transport is powered by the actin-dependent motor myosin-V. Aggregation and dispersion of melanosomes were induced by the addition of melatonin and MSH, respectively.

The MSD of melanosomes in both stimulation conditions displayed a transition from a subdiffusive to a superdiffusive regime and the corresponding local slope of the MSD-versus-lag time was well described by Eq. (23), as shown in Fig. 4.

Four parameters characterize the behavior:  $\lambda$ ,  $\alpha$ ,  $\varepsilon$ , and  $\delta$ , where  $\lambda$  and  $\alpha$  are the power-law exponents of the internal and external noise correlation functions,  $\varepsilon$  is a parameter that measures the relative intensity between random forces, and  $\delta$  is associated with the residual value of the MSD as  $\tau \rightarrow 0$ .

The parameters  $\lambda$ ,  $\alpha$ ,  $\epsilon$ , and  $\delta$  obtained from the fitting shown in Fig. 4 are displayed in Table I. The power-law exponents  $\lambda$  and  $\alpha$  for both dispersing and aggregating

melanosomes were similar ( $\lambda \sim 0.97$  and  $\alpha \sim 0.58$ ). These values lead to an asymptotic exponent  $2\lambda - \alpha \sim 1.36$ , which is a signature of superdiffusion. Similar values were obtained in other systems [8,12,14,19]. On the other hand, the obtained value  $\alpha \sim 0.58$  implies a power spectrum of fluctuating motor forces that scales as  $\omega^{-0.42}$ , corresponding to a smoothing of discontinuities in instantaneous force pulses, as was stated in Refs. [8,12].

It is worth pointing out that our model enables us to discern the contributions of internal and external forces in a straightforward fashion. In a recent paper [12], Wilhelm described the dynamics of forced motions of magnetic microbeads in *Dictyostelium* cells in terms of a generalized Langevin equation. However, differently of what we have done in this paper, that model considers a single force term which included contributions of both thermal Brownian forces and driving forces generated by molecular motors. The force correlation function followed a power-law behavior with exponent 0.8, which is between the values of  $\lambda$  and  $\alpha$  obtained by us.

Melanosomes in cells expressing a dominant-negative myosin-V construct, and thus not being actively transported by myosin-V, show spontaneous motion and the MSD displays a subdiffusive behavior. In the absence of motors, Eq. (23) rewrites in terms of  $\lambda$  and  $\delta$  [Eq. (25)]. Interestingly, we obtained a value for  $\lambda \sim 0.94$ , very similar to the one obtained in the presence of active motors. This reinforces our hypothesis that  $\lambda$  is the parameter that characterizes the subdiffusive motion of the melanosomes in the cytoplasm. In both cases (i.e., in the presence and in the absence of active motors),  $\lambda$  is close to 1. Similar values were obtained for the subdiffusive anomalous exponent of tracer proteins in highly concentrated random-coil polymer and globular protein solutions that mimic the crowded conditions encountered in cellular environments [23], and for the anomalous diffusion of dextran polymers inside HeLa cells [21].

Typically, the anomalous diffusion regime is characterized by fitting an empirical power-law relationship to the MSD dependence with  $\tau$  of the form [13,31,39]:

$$\rho(\tau) = C + D^* \left(\frac{\tau}{\tau_0}\right)^{\gamma},\tag{26}$$

being  $D^*$  an effective diffusion coefficient and  $\gamma$  an exponent between 0 and 2. Notice that with this definition  $D^*$  has units of distance square.

Likewise, we can define the effective diffusion coefficient for a subdiffusive and a superdiffusive limit regimes in our model. In the first case, setting  $\varepsilon$ =0 and  $\gamma$ = $\lambda$  in (20), we have that

$$D_{sub}^* = \frac{(2\eta)^2}{\delta} \frac{1}{\Gamma(\lambda+1)}.$$
 (27)

In the superdiffusive limit, setting  $\gamma=2\lambda-\alpha$  in Eq. (20) we obtain

$$D_{sup}^* = \varepsilon \frac{(2\eta)^2}{\delta} K_{\lambda,\alpha}, \tag{28}$$

where we have used relation (24) to derive this expression.

Interestingly, these analytical expressions enable a quantitative interpretation of the physics underlying the transport processes, linking the macroscopic effective diffusion coefficient with microscopic parameters derived from the forces correlation functions.

In Ref. [13], we obtained the values of  $D^*$  by fitting the MSD-vs-time lag plot of each trajectory with the empirical expression given by Eq. (26). The distributions of  $D^*$  followed an exponential behavior, with mean values equal to  $2 \times 10^{-3}$  and  $4.4 \times 10^{-3}$   $\mu m^2$  for aggregation and dispersion conditions, respectively.

Using (28) and the parameters displayed in Table I to compute the analytical effective diffusion coefficient, we obtained  $(5\pm2)\times10^{-3}$  and  $(4\pm1)\times10^{-3}$   $\mu\text{m}^2$ , for aggregation and dispersion, respectively. These values are consistent with the ones obtained empirically.

For mutant cells, the computed analytical effective diffusion coefficient (27) is  $D^* = (0.85 \pm 0.1) \times 10^{-3} \ \mu\text{m}^2$ , which is in agreement with the one obtained in Ref. [13]  $(\sim 1 \times 10^{-3} \ \mu\text{m}^2)$ .

Moreover, the correlation function (9) allows obtaining an estimation for the magnitude of the "global" force exerted by the motors  $F_{mot}$ , which is given by

$$F_{mot} \approx \sqrt{\frac{\Lambda_0}{\Gamma(1-\alpha)}}$$
 (29)

Using definitions (18) and (24),  $\Lambda_0$  rewrites as

$$\Lambda_0 = \varepsilon \frac{\delta}{(2n)^2} (2k_B T)^2. \tag{30}$$

Taking  $k_BT=4\times10^{-3}$  pN  $\mu$ m,  $(2\eta)^2=\rho_o$ , and  $\epsilon$ ,  $\delta$ , and  $\alpha$  from Table I, we obtain  $F_{mot}\sim8\pm4$  pN and  $\sim16\pm6$  pN, for dispersion and aggregation, respectively, which are in the order of the reported stall force of a single myosin-V [40]. The values obtained for aggregation and dispersion are similar within the experimental error, since the stall forces for myosin in these stimulation conditions are expected to be the same.

#### V. SUMMARY

The violation of the fluctuation-dissipation theorem in living cells has been observed in recent studies [8,11,12,14]. Motor proteins, which are force generators in cells, cannot only modify the viscoelastic response of the cytoplasm, as is the case of cytoskeleton motors, but they also generate non-thermal random forces which drive the system out of equilibrium [11]. The action of motors is generally reflected in a transition from a subdiffusive to a superdiffusive behavior of the displacement of particles within the cells [8].

To investigate how the action of molecular motors affects the transport in living cells, we study the motion of melanosomes driven by myosin-V along the *F*-actin network in *Xenopus laevis* melanocytes.

Differently to other microrheology experiments reported before [7,8,11,12,14], we adopt a different approach which uses a generalized Langevin equation that explicitly considers the collective action of the molecular motors and the presence of experimental/biological noise. The analytical solution of the model is obtained and an expression for the slope of the MSD vs the time lag is derived explicitly. The comparison between the model solution and the experimental data is straightforward, and involves four parameters. Two of these parameters are the exponents of the thermal and non-thermal forces correlation functions, while the others are related with the relative intensity between both kind of forces and the residual value of the MSD.

On one hand, the model predicts the observed crossover between subdiffusive to superdiffusive regimes, as well as it gives good estimates for the *in vivo* motor forces. On the other hand, the proposed method enables us to determine a link between the macroscopic effective diffusion coefficient and the parameters in the microscopic scale.

We believe that this theoretical approach can be used to describe the dynamics of intracellular transport of different cargoes in other living cells.

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#### APPENDIX: MATERIALS AND METHODS

#### 1. Melanophore cell culture and transfection

Immortalized *Xenopus laevis* melanophores were cultured as described in Ref. [35]. In order to track the movement of individual organelles, the number of melanosomes in the cell was reduced by treatment with phenylthiourea [5].

To study transport along actin filaments, the cells were incubated at 0 °C for 30 min with 10  $\mu$ M nocodazole to depolymerize microtubules [28].

Melanophores were stimulated for aggregation or dispersion with 10 nM melatonin or 100 nM MSH, respectively. The samples were observed between 5 and 15 min after stimulation. All the measurements were performed at 21 °C.

Cells were transfected using the FuGENE 6 transfection reagent (Boehringer Mannheim Corp.) following the vendor's protocols with a plasmid encoding a green fluorescent protein-tagged myosin-V tail. Expression of this plasmid results in a dominant-negative inhibition of myosin-V driven melanosome transport [41]. The plasmid was a kind gift of Doctor Vladimir Gelfand (Northwestern University, Chicago, IL).

## 2. Samples preparation for imaging

For microscopy measurements, cells were grown for 2 days on 25-mm round coverslips placed into 35-mm plates in 2.5 ml of the medium. Before observation, the coverslips were washed in serum-free 70% L-15 medium and mounted in a custom-made chamber specially designed for the microscope.

#### 3. Tracking experiments

Single particle tracking experiments of melanosomes moving along actin filaments in wild-type cells were carried out in a Zeiss IM 35 microscope adapted for SPT using a 63x oil-immersion objective (numerical aperture=1.25) under illumination with a tungsten-halogen lamp. A CCD camera (PixelVision, Oregon, USA) was attached to the video port of the microscope for imaging the cells. Movies were registered at a speed of 14 frame/s.

Tracking experiments of melanosomes in transfected cells were carried out in an Olympus IX70 microscope using a 60x water-immersion objective (numerical aperture=1.2). A

- cMOS camera (Pixelink, Ottawa, Ontario, Canada) was attached to the video port of the microscope for imaging the cells at a speed of 50 frame/s.
- Trajectories of melanosomes were recovered with 2 nm precision and 10 ms temporal resolution from the movies registered as described above using the pattern-recognition algorithm described in [3]. This algorithm is included in the program Globals for Images developed at the Laboratory for Fluorescence Dynamics (UCI, Irvine, CA). The program, which also contains some of the tools used for trajectory analysis, can be downloaded from the Laboratory for Fluorescence Dynamics website (www.lfd.uci.edu).
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