

Supplementary Information for “A tug of war between filament treadmilling and myosin induced contractility generates actin cortex”

Qin Ni,¹ Kaustubh Wagh,² Arpita Upadhyaya,^{2,3*} Garegin A. Papoian,^{2,4*}

¹Department of Chemical and Biomolecular Engineering

²Department of Physics

³Institute for Physical Science and Technology

⁴Department of Chemistry and Biochemistry

University of Maryland, College Park, MD, USA

*To whom correspondence should be addressed; E-mail: arpita@umd.edu; gpapoian@umd.edu.

1 Computational Model Details

The simulations presented in this work were carried out via MEDYAN (Mechanochemical Dynamics of Active Networks). MEDYAN is an open-access mechanochemical simulation platform (available at www.medyan.org) for active matter (*1*). We employed MEDYAN to model stochastic reaction-diffusion, mechanical interaction, and mechanochemical dynamics of active cytoskeletal networks.

In MEDYAN, the simulation space is divided into a solution phase and a polymer phase. All diffusing molecules (DM) are dissolved in the solution phase, including G-actin monomers, unbound formin, unbound non-muscle myosin II (NMII), and unbound crosslinker molecules. To reduce the computational cost without losing too much spatial information, the solution phase is discretized into many linear compartments. The dimension of the compartment is carefully

chosen based on the Kuramoto length of actin, which is the mean-free path that molecules are expected to diffuse before undergoing their next reaction (2). In the spherical network, compartments are $500 \text{ nm} \times 500 \text{ nm} \times 500 \text{ nm}$ cubes. In thin oblate networks, the compartment dimension in the Z-axis is reduced to 400 nm (the mechanical boundary is 200 nm as restricted by the boundary repulsion potential, see Section 1.2 for details). Diffusing molecules are assumed to be well-mixed within each compartment without specific spatial locations, and the transport of molecules between compartments is modeled as a diffusion reaction within the solution phase.

On the other hand, polymeric filaments, bound formin, bound NMII, and bound crosslinkers comprise the polymer phase that lays over the solution phase. This phase accounts for the mechanical modeling of boundary repulsion, steric interactions, bending and stretching of filaments, as well as the stretching of linkers and motors. When polymerization, nucleation, NMII or crosslinker binding reactions occur, diffusing molecules transfer from the solution phase into the polymer phase. Depolymerization, filament destruction, NMII or crosslinker unbinding reactions will release molecules from the polymer phase to solution phase.

In the following sections, we will further discuss details of our mechanical models, chemical reaction-diffusion models, mechanochemical coupling, and the simulation protocol.

1.1 Mechanical models

Unlike the traditional bead-spring model, the semi-flexible filaments are represented as connected cylinders. The equilibrium length (under zero force) of each cylinder elements varies from 2.7 nm (1 actin monomer) to a maximum of 108 nm (40 actin monomers). Addition of each actin monomer would increase the length of the first or last cylinders by 2.7 nm , and vice versa. Polymerization will create a new cylinder if the cylinder has reached its maximum length. Filaments have a very large aspect ratio, i.e., the persistence length of a filament ($\sim 20 \mu\text{m}$) is

much larger than its diameter ($\sim 10nm$). Thus, it is reasonable to ignore the radial stretching/compression and only allow the axial stretching/compression of a cylinder, which is written as

$$U_{filament}^{str} = \frac{1}{2} K_{filament}^{str} (l_f - l_{f,0})^2.$$

l_f is the actual length of cylinder under force, and $l_{f,0}$ is the equilibrium length based on the number of actin monomers on this cylinder (each monomer is 2.7nm). Radial filament deformation is modeled as bending between two connected cylinders:

$$U_{filament}^{bending} = K_{filament}^{bending} (1 - \cos(\theta - \theta_0)),$$

where θ is the angle between the two consecutive cylinders under force, while θ_0 is the equilibrium angle that is set to be 0.

A novel volume exclusion potential is implemented to prevent cylinders overlapping, which is written as

$$U^{Vol} = \iint_{l_i, l_j} \delta U | \vec{r}_i - \vec{r}_j | dl_i dl_j,$$

where $\delta U | \vec{r}_i - \vec{r}_j | = 1 / | \vec{r}_i - \vec{r}_j |^4$ is the pair potential between two points located on the two interacting cylinders. \vec{r}_i and \vec{r}_j are the distances between any two points along the cylinder i and j , respectively. This potential can provide a steep enough volume exclusion effect while remain analytically solvable.

Bound NMII and linkers are modeled as harmonic springs, and the stretching energy is written as

$$U_{NMII/linker}^{str} = \frac{1}{2} K_{NMII/linker}^{str} (l_{NMII/linker} - l_{NMII/linker,0})^2.$$

$l_{NMII/linker,0}$ is the equilibrium length of a linker, which are initialized when a linker /NMII binding reaction occurs as the distance between the paired binding site. $l_{NMII,0}$ is reset every time a motor walking reaction occurs.

Table SI-1: Mechanical parameters.

Names	Parameters	References
Cylinder stretching	$K_{filament}^{str} = 100pN/nm$	1
Cylinder bending	$K_{filament}^{bending} = 672pN \cdot nm$	3
Filament volume exclusion	$K_{vol} = 10^5pN/nm^4$	1
Linker stretching	$K_{linker}^{str} = 8pN/nm$	4
NMII stretching	$K_{NMII}^{str} = 2.5pN/nm$ per head	5
Boundary repulsion	$\epsilon_{boundary} = 100pN \cdot nm$	This work

In order to confine all the filaments within the simulation boundary, an exponential boundary repulsion potential is implemented. In the thin oblate system, the actual height of the network is set to be 400 nm, and the diameter to 4000 nm. However, filaments would occasionally move out of the mechanical boundary due to rapid treadmilling, leading to simulation failures. To prevent this, we shift the boundary barrier slightly inside the network by a_0 , and the exponential boundary repulsion is written as

$$U^{boundary} = \epsilon_{boundary} e^{-(d-a_0)/\lambda},$$

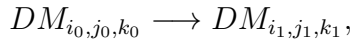
where $\epsilon_{boundary} = 100pN \cdot nm$ is the repulsive energy constant, d is the distance between boundary and filament element, and $\lambda = 2.7nm$ is the screening length. The boundary shifting factor a_0 is chosen to be 100 nm based on experience. The existence of a_0 restricts the effective network boundary to height = 200 nm and diameter = 3800 nm.

The mechanical model parameters can be found in Table SI-1.

1.2 Chemical models

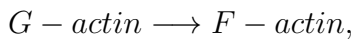
The chemical engine of MEDYAN is powered by Next Reaction Method (NRM) (6), which is a variant of the Gillespie algorithm (7). Overall, the NRM stochastically solves the chemical Master Equation by generating a trajectory of chemical events. In this work, we simulated the following chemical reactions: diffusion, filament polymerization, filament depolymerization, filament nucleation, destruction of filaments, binding of myosin motors and linkers, and motor walking.

The diffusion of molecules is modeled as a single molecule transfer process between neighboring compartments, which follows our stochastic chemical reaction protocol as



where a diffusing molecule (DM) originally located in compartment i_0, j_0, k_0 is transferred to a neighboring compartment i_1, j_1, k_1 . The copy number of this diffusing molecule species is decreased by 1 in compartment i_0, j_0, k_0 and is increased by 1 in compartment i_1, j_1, k_1 .

Actin filament (F-actin) polymerization and depolymerization occur at both barbed end (BE) and pointed end (PE) of a filament. These reactions are written as



It should be noted that G-actin is dissolved in the solution phase, while F-actin is in the polymeric phase.

The nucleation reaction is presented as a two-step reaction based on the mechanism of formin nucleation (12, 13):

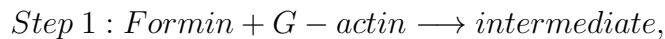
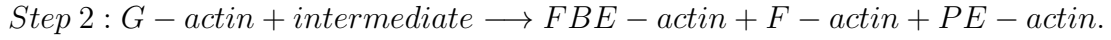
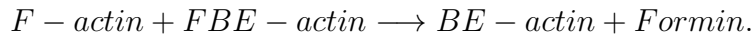


Table SI-2: Parameters for diffusion and reactions.

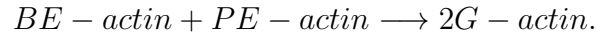
Names	Parameters	References
Diffusion	$D_{actin,arp2/3,CP} = 20\mu M^2/s$	2
Actin	$k_{on}^{BE} = 11.6 - 34.8(\mu M \cdot s)^{-1}$ $k_{on}^{PE} = 1.3(\mu M \cdot s)^{-1}$ $k_{off}^{BE} = 1.4s^{-1}$ $k_{off}^{PE} = 0.8 - 2.4s^{-1}$	8 and this work
Destruction	$k_{destruction} = 1.0 - 1.9s^{-1}$	This work
Nucleation	$k_{nu} = 0.005s^{-1}$	[Turnover]
Formin dissociation	$k_{off}^{formin} = 0.01s^{-1}$	9
Alpha-actinin	$k_{on}^{\alpha} = 0.7(\mu M \cdot s)^{-1}$ $k_{on}^{\alpha} = 0.3s^{-1}$	10
NMII head binding	$k_{on}^M = 0.2s^{-1}$ $k_{on}^M = 1.7s^{-1}$	11 1



The intermediate is an arbitrary molecule that consists of a formin and a G-actin molecule. We assume step 1 is the rate-limiting step and step 2 is a fast step, thus this intermediate would rapidly react with a G-actin molecule and become a short filament consisting of one F-actin molecule at the pointed end (PE-actin), a regular F-actin molecule, and another F-actin molecule at the formin bound barbed end (FBE-actin). For simplicity, polymerization and depolymerization at FBE are the same as regular barbed end reactions. Formin can dissociate from a filament, which releases a formin molecule into the solution phase and creates a regular F-actin barbed end (BE-actin) on that filament:

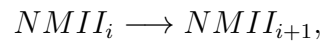


Since new filaments are constantly created by nucleation,, the filament destruction process is required to establish a steady state which maintains a constant total number of filaments. The destruction reaction occurs exclusively when a filament has only two F-actin molecules (a BE-actin and a PE-actin), which destroys this filament and releases two diffusing G-actin molecules as



The binding reactions of myosin motors and linkers are carried out with a slightly different protocol. Firstly, the system will search for all possible binding site pairs on actin filaments and stochastically choose one for binding reaction. The two binding sites of a pair must be located at different filaments. The distance between the two binding sites ranges from 175-225 nm for NMII mini filament (14), and 30-40 nm for alpha-actinin crosslinker (15). After the binding site pair is determined, the binding reaction convert a diffusing motor or linker to a bound motor or linker with two ends attaching to the two binding sites, creating a mechanical linkage. This linkage vanishes when an unbinding reaction occurs, releasing the motor or linker to the diffusing pool. It should be noted that NMII mini filament is an ensemble of 15-30 myosin heads (16), and we model the entire ensemble as a whole. To take the variation of the number of myosin heads into account, the number of myosin heads of each NMII mini filament is chosen stochastically for each reaction, and the reaction rate for each NMII binding event is then scaled by the number of myosin heads. .

In an active cytoskeleton, myosin motors consume energy from ATP hydrolysis and actively walk along filaments, which is one of the most important sources of contractile force generation. In MEDYAN, a motor stepping reaction is implemented to mimic this effect. For a bound NMII, the stepping reaction is written as



where i and $i + 1$ are the NMII locations on the filament before and after walking. NMII is a barbed end walking motor, thus $i + 1$ represents the next binding site towards the barbed end.

Parameters for diffusion and chemical reactions can be found in Table SI-2.

1.3 Mechanochemical models

Many cytoskeletal reactions, including actin polymerization, myosin motor binding and stepping, and linker binding, are mechanosensitive. To capture this feature, MEDYAN implements mechanochemical models that explicitly allow force-dependent chemical reaction rates.

Table SI-3: Mechanochemical dynamic rate parameters.

Names	Parameters	References
Characteristic polymerization force	$F_{poly,0} = 1.5pN$	17
Characteristic linker unbinding force	$F_{linker,unbind} = 17.2pN$	18
NMII duty ratio	$\rho = 0.1$	11
NMII stall force	$F_{stall} = 12.62pN$ per head	19
Tunable parameters	$\beta = 0.2$ $\gamma = 0.05pN^{-1}$ $\xi = 0.1$	1

The effect of boundary force on filament polymerization is described by the Brownian Ratchet model (20), which models the force sensitive polymerization rate k_{poly} as:

$$k_{poly} = k_{poly}^0 \cdot \exp(-F_{ext}/F_{poly,0}),$$

where k_{poly}^0 is the bare polymerization rate under zero external force, F_{ext} is the boundary repulsive force exerted on the filament ends, and $F_{poly,0}$ is the characteristic polymerization force

based on the thermal energy and the size of actin monomers.

We used a simple exponential equation to model the slip bond property of alpha-actinin crosslinker:

$$k_{linker,unbind} = k_{linker,unbind}^0 \cdot \exp(F_{linker,stretching}/F_{linker,unbind}),$$

where $k_{linker,unbind}^0$ is the unbinding rate constant under zero external force, and $F_{linker,unbind}$ is the characteristic unbinding force of alpha-actinin. $F_{linker,stretching}$ is the stretching force on the linker, while a compressive force on the linker does not trigger the slip bond.

In this work, we model NMII binding as a catch bond, as adapted from the Parallel Cluster Model (19), such that the force loaded on NMII can reduce its unbinding rate constant:

$$k_{NMII,unbind} = \frac{\beta \cdot k_{NMII,unbind}^0}{N_{heads}} \cdot \exp\left(\frac{-F_{ext}}{N_{heads} \cdot F_{NMII,unbind}}\right),$$

where β is a tunable parameter, $k_{NMII,unbind}^0$ is the unbinding rate constant under zero force, F_{ext} is the total stretching force applied on the NMII, and N_{heads} is the number of NMII heads.

The NMII walking rate is also mechanochemically sensitive and can be modeled with a Hill type force-velocity relation:

$$k_{walk} = k_{walk}^0 \cdot \frac{F_{stall} - F_{ext}/N_{heads}}{F_{stall} + F_{NMII,pulling}/(N_{heads} \cdot \xi)},$$

where F_{stall} is the stall force of a single NMII head, $F_{NMII,pulling}$ is the pulling force on NMII in the opposite direction of walking movement, and ξ is a tunable parameter.

The mechanochemical model parameters can be found in Table SI-3.

1.4 Simulation protocol

The relaxation time for local deformations of actin networks (21) is much shorter than the timescale of typical chemical events such as motor stepping (11) or filament polymerization (8),

thereby creating a significant separation of timescales. Hence, the mechanical equilibrium process can be viewed as a pseudo-adiabatic process that can be separated from chemical reactions. Based on this hypothesis, the simulation can be carried out in the following steps:

1. Chemical reactions occur that evolve the time of the system stochastically.
2. Pausing chemical reactions when the time step reaches a preset value, which is 10 ms in this work. The system then mechanically minimizes the total energy.
3. Reaction rates are updated based on the tension acting on NMII/linkers and load force acting on actin filament barbed ends after mechanical minimization.
4. Step 1 is repeated based on the updated reaction rates.

This protocol is iterated until we reach 2000 seconds of simulation time, or until we reach the wall time limit on the Deephought2 High-Performance Computing cluster at University of Maryland, College Park, whichever comes first.

2 Defining treadmilling rate and treadmilling inhibition simulation setups

Although treadmilling in cells is a complex system that involves hundreds of reactions (22, 23), it is simplified to four reactions in this work by considering polymerization and depolymerization at both barbed ends and pointed ends. When a steady state is established, the net barbed end growth rate will equal the net pointed ends reduction rate (averaged over the system), maintaining a constant average filament length. Therefore, we can define a kinetic steady state for treadmilling by monitoring the average filament length of the network as shown in Fig. SI-1. We found that such a kinetic steady state could be established after 1000s in all conditions, and at this state, the average barbed end elongation rate is almost the same as the average pointed end shrinkage rate. Hence, we quantify the average treadmilling rate $\langle r_{TM} \rangle$ as the average barbed end elongation rate after 1000s.

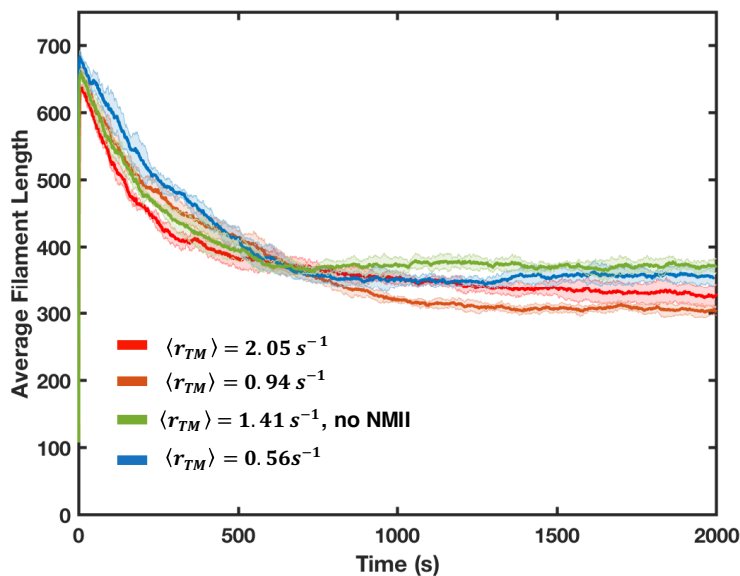


Figure SI-1: Average filament length as a function of time are shown. Shaded color represents the standard deviation of mean (5 run per condition).

While the treadmilling rate is an elegant and robust way of quantifying the speed of actin network assembly, it is extremely hard to measure *in vivo*. An alternative way to quantify the speed of actin network remodeling is to measure the turnover timescale, which has been widely studied via an experimental technique called Fluorescent Recovery After Photobleaching (FRAP). To compare with experiments, in our simulation we used a method mimicking the FRAP to calculate the turnover half-time ($t_{1/2}$, the time required for a network to reach 50% turnover) as developed in our previous work (13), and we obtain $t_{1/2} \sim 168s$ for the slowest treadmilling condition, and $t_{1/2} \sim 48s$ for the most rapid treadmilling case. It should be noted that our longest $t_{1/2}$ is similar to the turnover timescale of some reconstituted networks (24), and our shortest $t_{1/2}$ is comparable to that of *in vivo* actin cortices (25). The details of turnover half-time measurement in MEDYAN and how it is related to treadmilling has been discussed in depth in a prior computational study (13).

We utilized kinetic parameters measured *in vitro* (8) as the baseline to assemble the slow

treadmilling networks. To explore suitable parameters for rapidly treadmilling networks, we looked into the effects of formin and ADF/cofilin. An earlier work (26) has shown that the presence of formin can boost the polymerization rate at the barbed end several-fold over the baseline. For simplicity, we imitated this effect by increasing the barbed end polymerization rate constant (k_{on}^{BE}). ADF/cofilin can also promote treadmilling by severing filaments. Importantly, the fragment that contains the pre-existing pointed end is very unstable and would undergoes rapid disassembly (24). This observation allows us to mimic the effect of ADF/cofilin by simply increasing the depolymerization at the pointed end (k_{off}^{PE}). For example, we increase the k_{on}^{BE} and k_{off}^{PE} to three-fold in the actin ring network as shown in Fig. 1a-c ($\langle r_{TM} \rangle = 2.05s^{-1}$).

It is difficult to tune the polymerization and depolymerization rate *in vivo*. Thus, we indirectly perturb filament assembly and disassembly in T cells by treating them with Latrunculin A (LatA). Earlier works have shown that LatA affects filament treadmilling in two ways: 1) it sequesters G-actin and 2) it accelerates the phosphate release from ADP-Pi-actin thereby reducing filament polymerization while increasing depolymerization at both ends (27–29). To simulate such effects in the actin ring perturbation simulations, we explore a parameter space that mimicked the effect of LatA treatment: we disrupted r_{TM} by reducing the filament polymerization rate and increasing the depolymerization rates. In the weak inhibition case, we decreased k_{on}^{BE} to $11.6(\mu M \cdot s)^{-1}$, increased k_{off}^{BE} to $2.1s^{-1}$, and maintained k_{off}^{PE} at $2.4s^{-1}$. In the strong inhibition case, k_{on}^{BE} was decreased to $3.48(\mu M \cdot s)^{-1}$, k_{off}^{BE} was increased to $11.2s^{-1}$, and k_{off}^{PE} was increased to $4.8s^{-1}$. In all simulations, pointed end polymerization rate was set to be constant at $1.3(\mu M \cdot s)^{-1}$.

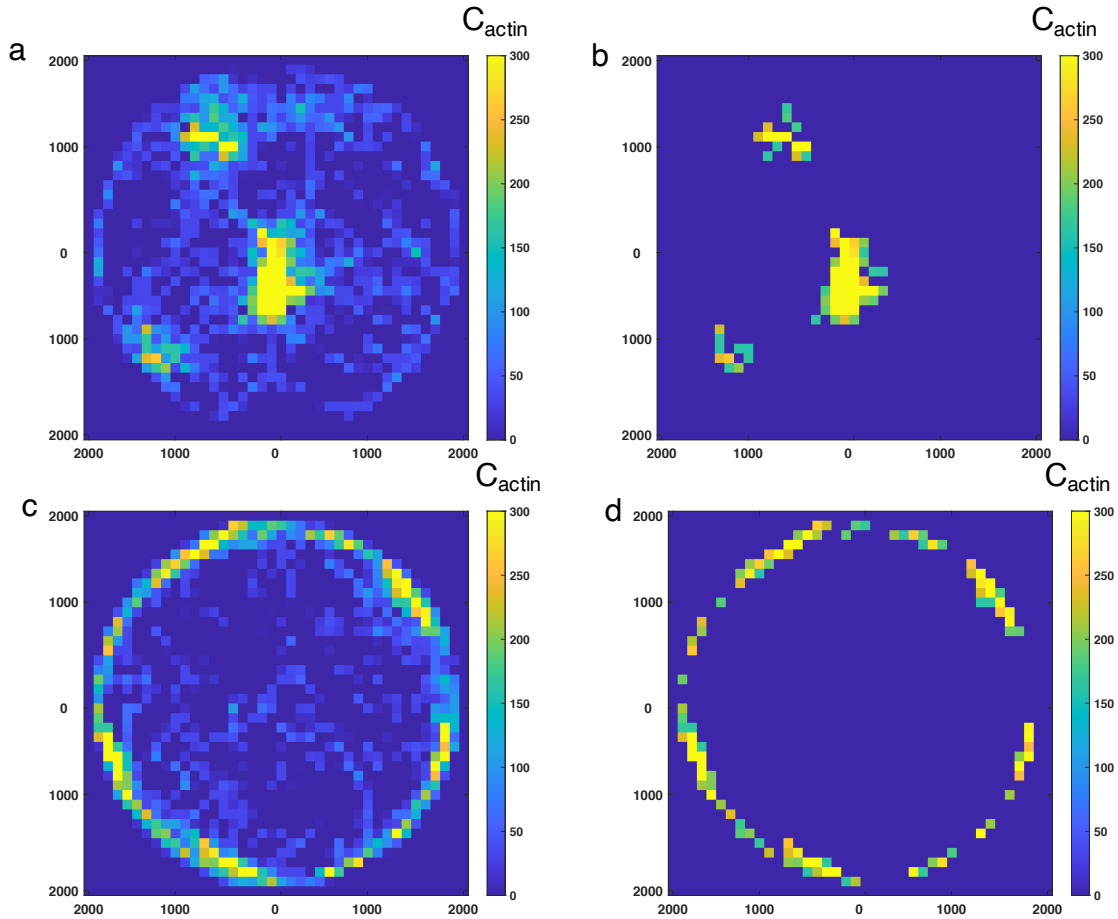


Figure SI-2: Heatmap showing a pixellated representation of the local F-actin concentration in 100nm x 100nm bins for (a) a cluster-like network as shown in Fig.1a-c ($\langle r_{TM} \rangle = 0.56s^{-1}$), and (b) a ring-like network as shown in Fig.1a-c ($\langle r_{TM} \rangle = 2.05s^{-1}$). (b and d) are the same heatmaps as (a) and (c), respectively, but only contains bins that exceed a threshold concentration of 160 μM .

3 Calculation of local actin concentration for clusters and rings

In this work, we used a density-based clustering method to define regions that contain actin clusters and rings, and calculated the local F-actin concentration within these regions. We first

a generated pixelated map by dividing the network into $100nm \times 100nm$ bins and calculated the F-actin concentration within each bins (Fig. SI-2a). We then grouped connecting bins with concentration higher than a threshold ($160 \mu M$) into clusters (Fig. SI-2b). Clusters with size less than 4 bins were ignored. The local actin concentration within clusters was calculated as the average F-actin concentration of these clusters. The local actin concentration within actin rings is calculated using the same method (Fig. SI-2c-d).

References

1. K. Popov, J. Komianos, G. A. Papoian, *PLoS Computational Biology* **12**, 1 (2016).
2. L. Hu, G. A. Papoian, *Biophysical Journal* **98**, 1375 (2010).
3. A. Ott, M. Magnasco, A. Simon, A. Libchaber, *Physical Review E* **48** (1993).
4. B. A. Didonna, A. J. Levine, *Physical Review E - Statistical, Nonlinear, and Soft Matter Physics* **75**, 1 (2007).
5. A. Vilfan, T. Duke, *Biophysical Journal* **85**, 818 (2003).
6. M. A. Gibson, J. Bruck, *The Journal of Physical Chemistry A* **104**, 1876 (2000).
7. D. T. Gillespie, *Journal of Physical Chemistry* **81**, 2340 (1977).
8. I. Fujiwara, D. Vavylonis, T. D. Pollard, *Proceedings of the National Academy of Sciences* **104**, 8827 (2007).
9. M. Fritzsche, C. Erlenka mper, E. Moendarbary, G. Charras, K. Kruse, *Science Advances* **2**, e1501337 (2016).
10. D. Wachsstock, W. Schwartz, T. Pollard, *Biophysical Journal* **65**, 205 (1993).

11. M. Kovács, F. Wang, A. Hu, Y. Zhang, J. R. Sellers, *Journal of Biological Chemistry* **278**, 38132 (2003).
12. M. Pring, M. Evangelista, C. Boone, C. Yang, S. H. Zigmond, *Biochemistry* **42**, 486 (2003).
13. Q. Ni, G. A. Papoian, *Cytoskeleton* **76**, 562 (2019).
14. T. D. Pollard, *Journal of Cell Biology* **95**, 816 (1982).
15. R. K. Meyer, U. Aebi, *Journal of Cell Biology* **110**, 2013 (1990).
16. A. B. Verkhovsky, T. M. Svitkina, G. G. Borisy, *The Journal of cell biology* **131**, 989 (1995).
17. M. J. Footer, J. W. Kerssemakers, J. A. Theriot, M. Dogterom, *Proceedings of the National Academy of Sciences of the United States of America* **104**, 2181 (2007).
18. J. M. Ferrer, *et al.*, *Proceedings of the National Academy of Sciences of the United States of America* **105**, 9221 (2008).
19. T. Erdmann, P. J. Albert, U. S. Schwarz, *Journal of Chemical Physics* **139** (2013).
20. C. S. Peskin, G. M. Odell, G. F. Oster, *Biophysical Journal* **65**, 316 (1993).
21. T. T. Falzone, S. Blair, R. M. Robertson-Anderson, *Soft Matter* **11**, 4418 (2015).
22. B. Bugyi, M.-F. Carlier, *Annual review of biophysics* **39**, 449 (2010).
23. C. S. Floyd, C. Jarzynski, G. A. Papoian, *New Journal of Physics* (2017).
24. P. M. McCall, F. C. MacKintosh, D. R. Kovar, M. L. Gardel, *Proceedings of the National Academy of Sciences of the United States of America* **116**, 12629 (2019).
25. G. Salbreux, G. Charras, E. Paluch, *Trends in Cell Biology* **22**, 536 (2012).

26. D. R. Kovar, E. S. Harris, R. Mahaffy, H. N. Higgs, T. D. Pollard, *Cell* **124**, 423 (2006).
27. H. F. Lodish, *Molecular cell biology* (W.H. Freeman, 2000).
28. E. G. Yarmola, *Journal of Biological Chemistry* **275**, 28120 (2000).
29. I. Fujiwara, M. E. Zweifel, N. Courtemanche, T. D. Pollard, *Current Biology* **28**, 3183 (2018).