Energy-based Modelling of Single Actin Filament Polymerisation Using Bond Graphs

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Abstract

Energy-based modelling of single actin filament polymerisation and force generation is investigated using the bond graph approach. It is shown that the **TF** (transformer) bond graph component provides a practical, and conceptually simple, alternative to the Brownian ratchet approach of Peskin, Odell, Oster & Mogilner. Three cases are examined: rigid filament normal to a surface; rigid filament at an angle to a surface and a flexible filament. The latter two cases correspond to incremental additions to the bond graph of the first case. In the first case, the explicit formula derived from the bond graph is identical to that derived by the Brownian ratchet approach. Energy flows are fundamental life; for this reason, the energy based approach is utilised to investigate the power transmission by the actin filament and its corresponding efficiency.

The bond graph model is fitted to experimental data by adjusting model physical parameters.

1   Introduction

Actin filament mechanics and dynamics govern cell and tissue physiology throughout life. Actin dynamics and organisation play an integral role in the formation of tissues at cell-cell adhesions [1], play a central role in cell division and growth [2], contribute to cell stiffness [3] and interact with myosin to generate contractile forces in motile and muscle cells.

The growth and breakdown of actin filaments underlies actin dynamics. The rate of actin polymerisation is tightly regulated by an interconnected network of chemical reactions, and mechanical forces. In a seminal paper on actin-based cellular motility, Peskin et al. [4] introduce the concept of a *Brownian ratchet* to "present a mechanical picture of how polymerising filaments can exert mechanical forces". The mechanism provides a physical and thermodynamically consistent explanation of how the Gibbs free energy of actin polymerisation is converted into a mechanical pushing force at the leading edge. The Brownian ratchet approach has been embellished in a number of ways to provide a better fit to experimental data and remains at the core of current understanding of how actin polymerisation exert mechanical forces [5–7]. An elegant summary of the work of Peskin, Odell, Oster & Mogilner is given by Pollard [8] who highlights three key papers: [4] which introduces the concept of the Brownian ratchet in the context of a rigid actin filament touching a membrane at a right-angle to the membrane surface; [9] which extends the theory to flexible filaments touching the membrane at an arbitrary angle; and [10] which further extends the theory to include tethered filaments and capped filaments.

The concept of *energy transduction* as a basis for modelling the interface between chemistry and mechanics has a long history [11, 12]. Bond graphs were introduced by Paynter [13, 14] to model *energy* transduction within and between different physical domains including electrical, mechanical and hydraulic; a number of introductory texts are available [15–17]. Bond graph models of chemical reaction networks were introduced in the seminal papers of Oster et al. [18, 19], and later extended in the context of Systems Biology [20–27]. A recent introduction and overview of the bond graph approach to systems biology is available [28] and and a short introduction to bond graph modelling is given as [22, Appendix A].

In this study, we present a novel formulation of single actin filament polymerisation within the bond graph modelling framework. Moreover, the bond graph derived model of actin polymerisation is compared with the Brownian ratchet model. We illustrate that following a bond graph approach produces a force-velocity relationship identical in form to that of the Brownian ratchet although the bond graph model does not assume a Brownian ratchet mechanism.

The *transformer* **TF** the bond graph component which provides the link between different physical domains; in the context of systems biology, the **TF** component has been used to link the chemical and electrical domains [23, 29, 30] and to represent stoichiometry [28]. This paper suggests that the **TF** component can be utilised in place of the Brownian ratchet as the link between the chemical and mechanical domains. In particular, the chemomechanical behaviour of a single actin filament is analysed using the bond graph **TF** approach and the corresponding force-velocity curve is derived for rigid and flexible filaments. § 2 considers the rigid case normal to the surface and the result is explicitly compared with the Brownian ratchet results of Peskin et al. [4]. § 3 considers the power transfer and efficiency of the energy transduction. § 4 extends the bond graph model to the non-normal case and § 5 further extends the model to the flexible case. Comparison with the experimental data of Li et al. [7] (corresponding to that of Bieling et al. [6]) is presented in § 6.

The Python/Jupyter notebook used to generate the Figures is available at: `https://github.com/gawthrop/ActinRod24`.

2   Bond Graph model of chemomechanical transduction

This section considers a bond graph approach to chemomechanical transduction in the context of a rigid filament of actin impinging on a membrane normal to the membrane surface. A recent introduction and overview of the bond graph approach to systems biology is available [28] and and a short introduction to bond graph modelling is given as [22, Appendix A].

Figure 1 shows the corresponding bond graph model. The two species components **Ce**:A and **Ce**:B, together with the reaction component **Re**:r represent a chemical reaction [20, 28] where **Ce**:A and **Ce**:B represent complexes [24] (in the Chemical Reaction Network sense [31]). The **C**:F component represents the force exerted by the membrane on the filament. The bond graph **TF** component is the key to the transformation between physical domains in general and the chemical and mechanical domains in particular; this is considered in § 2.1.

It is shown in § 2.2 that that the bond graph approach yields an identical force/velocity equation to that derived by the Brownian ratchet approach of Peskin et al. [4].

2.1   The **TF** component and modulus

The **TF** component provides the interface between two different physical domains; in the context of systems biology, the **TF** component has been used to link the chemical and electrical domains [23, 29, 30]. The **TF** component is used here to link the chemical and mechanical domains. Although the chemical and mechanical domains differ and correspond to different physical units, each has the same notion of energy measured in Joules (J). The bond graph approach

Figure 1: Bond Graph model – rigid filament normal to surface.

(a) The chemical reaction A $\rightleftharpoons$ B describes the biochemistry of the growing actin filament. The reaction rate v $(\mathrm{mol\,s^{-1}})$ also provides the energy flow $(\mathrm{J\,s^{-1}})$ from the chemical to the mechanical domain – chemomechanical transduction. In this section, the actin filament is normal to the membrane. The velocity of the actin tip touching the membrane V $(\mathrm{m\,s^{-1}})$ depends on the length and number of monomers added to the polymer per second and is proportional to reaction rate v: $V = mv$. The constant m has units of $\mathrm{m\,mol^{-1}}$ and is derived in § 2.1. The growth of the filament is resisted by a force F (N).

(b) The bond graph is a precise representation of the energy flows implied by the conceptual diagram where $\rightharpoondown$ corresponds to mechanical or chemical energy flow $(\mathrm{J\,s^{-1}})$. The bond graph **TF**:m component represents the chemomechanical transduction with modulus m $(\mathrm{m\,mol^{-1}})$. Components **Ce**:A, **Ce**:B represent *chemical complexes* (mixtures of species) [24, 31] including ATP hydrolysis; **Ce**:A includes the Actin polymer with n units and the monomer; **Ce**:B includes the Actin polymer with $n + 1$ units. The bond graph component **Re**:r represents the chemical reaction. The *mechanical* spring **C**:F represents the force F (N) resisting the polymerisation of Actin and the corresponding tip velocity V $(\mathrm{m\,s^{-1}})$; this can be thought of as the mechanical equivalent of a chemical chemostat. The annotation in blue is purely explanatory and is not part of the bond graph.

assigns two key variables to each domain: an effort e and a flow f which have the property that their product is power $p = ef$ $\mathrm{J\,s^{-1}}$. Then chemical domain has chemical potential μ $\mathrm{J\,mol^{-1}}$ as effort and molar flow v $\mathrm{mol\,s^{-1}}$ as flow and the mechanical domain has force F N as effort and velocity V $\mathrm{m\,s^{-1}}$ as flow.

Because the **TF** component does not store or dissipate energy, it follows that both effort and flow are related by the same modulus m:

$$V = mv \tag{2.1}$$
$$\mu_F = mF \tag{2.2}$$

where μ_F is the chemical potential corresponding to the force F.

Equation (2.1) converts molar flow into mechanical velocity. As a mole of actin subunits contains N_A molecules where N_A $(\mathrm{mol^{-1}})$ is the Avogadro Constant, and assuming that the length of the Actin subunit is δ (m), it follows that the **TF** modulus m $(\mathrm{m\,mol^{-1}})$ is given by

$$m = N_A \delta \tag{2.3}$$

Because m appears in both equations (2.1) and (2.2) it follows that the modulus m derived from considering the flows can also be used to relate the chemical and mechanical efforts μ_F and F.

2.2 Comparison with the Brownian ratchet approach

From Figure 1, the net chemical potentials acting on the reaction component **Re**:r are $\mu_A - \mu_F$ and μ_B. Hence, the net reaction flow v is given the formula [20, 28]:

$$v = \kappa \left(e^{\frac{\mu_A - \mu_F}{RT}} - e^{\frac{\mu_B}{RT}} \right) = \kappa \left(e^{\frac{\mu_A - mF}{RT}} - e^{\frac{\mu_B}{RT}} \right) \tag{2.4}$$

where μ_A and μ_B are the chemical potential associated with complexes A and B, μ_F is given by Equation (2.2), F is the force associated with the mechanical spring and m is the transformer modulus. RT $(\mathrm{J\,mol^{-1}})$ is the product of the universal gas constant R $(\mathrm{J\,K^{-1}\,mol^{-1}})$ and the absolute temperature T (K).

The chemical potentials are given by

$$\mu_A = RT \ln K_A x_A \tag{2.5}$$
$$\mu_B = RT \ln K_B x_B \tag{2.6}$$

where K_A and K_B are thermodynamic constants for A and B [20].

Substituting (2.5) and (2.6) into (2.4) gives the molar flow rate:

$$v = \kappa \left(e^{\frac{-mF}{RT}} K_A x_A - K_B x_B \right) \tag{2.7}$$

Using equations (2.1) and (2.3), the mechanical velocity $V = mv$ is:

$$V = N_A \delta \kappa \left(e^{\frac{-N_a \delta F}{RT}} K_A x_A - K_B x_B \right) \tag{2.8}$$

Peskin et al. [4], equation (3) show that:

$$V = \delta \left(\alpha e^{-\omega} - \beta \right) \tag{2.9}$$

$$\text{where } \omega = \frac{F\delta}{k_B T} \tag{2.10}$$

k_B is the Boltzmann constant and T the absolute temperature. Equations (2.8) and (2.9) are identical if the following substitutions are made:

$$R = N_A k_B \tag{2.11}$$

$$\alpha = N_A \kappa K_A x_A \tag{2.12}$$

$$\beta = N_A \kappa K_B x_B \tag{2.13}$$

Equation (2.11) is the standard formula for the Gas Constant R; Equation (2.12) represents the polymerisation rate in monomers s^{-1} and Equation (2.13) represents the depolymerisation rate in monomers s^{-1}.

Thus the formula (2.8) obtained by bond graph arguments and the formula (2.9) obtained by statistical mechanics arguments by Peskin et al. [4] are identical.

2.3 Force-velocity curve

Name	Sm.	Value	Unit	Reference
Polymerisation rate	α	113	s^{-1}	Peskin et al. [4], Fig. 1
Depolymerisation rate	β	1.6	s^{-1}	Peskin et al. [4], Fig. 1
Monomer length	δ	2.7	nm	Peskin et al. [4], Fig. 1
Polymer length	L	30	nm	Mogilner and Oster [9], Fig. 2
Persistence length	λ	1	μm	Mogilner and Oster [9], Tab. 1
Contact angle	θ_0	90-54=36	degree	[7], equation (8)

Table 1: Numerical values from the literature

Using the $\alpha - \beta$ notation, Equation (2.8) becomes:

$$V = \delta \left(e^{\frac{-mF}{RT}} \alpha - \beta \right) \tag{2.14}$$

Setting $F = 0$ in equation (2.14), the velocity V_0 corresponding to no membrane force is given by:

$$V_0 = \delta \left(\alpha - \beta \right) \tag{2.15}$$

Using the values from Table 1:

$$V_0 = 0.30 \, \mu\text{m s}^{-1} \tag{2.16}$$

This value is similar to $V_0 = 0.2 \, \mu\text{m s}^{-1}$ quoted by Pollard and Borisy [32].

(a) Normalised   (b) Not normalised

Figure 2: Force-velocity curve. Using the values from Table 1, $\gamma = 4.26$ and the transformer modulus $m = 1.63 \times 10^{15}\,\mathrm{m\,mol^{-1}}$. The curves labelled $\gamma = 4.26$ are plotted for the values given in Table 1. The other two curves are for $\frac{\alpha}{4}$ ($\gamma = 2.87$) and 4α ($\gamma = 5.64$). (a) Normalised curves from Equation (2.27). (b) Without normalisation. This alternative representation emphasises that increasing the chemical potential α of the complex represented by **Ce**:**A** of Figure 1, and thus γ and the reaction potential Φ_0, increases both the stall force F_0 and the zero-force velocity V_0.

When the filament is stalled, $V = 0$, and the corresponding stall force F_0 is given by:

$$e^{\frac{-mF_0}{RT}}\alpha = \beta \tag{2.17}$$

$$\text{hence } F_0 = \gamma\frac{RT}{m} \tag{2.18}$$

$$\text{where } \gamma = \ln\frac{\alpha}{\beta} \tag{2.19}$$

Using the values from Table 1:

$$F_0 = 6.75\,\mathrm{pN} \tag{2.20}$$

$$\text{and } \gamma = 4.26 \tag{2.21}$$

This value for F_0 is within the range $F_0 = 7.7 \pm 1.3\mathrm{pN}$ quoted by Abraham et al. [33, Table 2].

Noting that the *chemical* reaction potential (affinity) Φ_c for the reaction is

$$\Phi_c = \mu_A - \mu_B = RT\ln\frac{\alpha}{\beta} = RT\gamma \tag{2.22}$$

it follows that the stall force F_0 can also be written as:

$$F_0 = \frac{\Phi_c}{m} \tag{2.23}$$

The force-velocity curve of Equation (2.14) can be simplified by using the expressions for V_0 and F_0 (2.15) and (2.18). In particular, equation (2.14) can be rewritten by normalising velocity

V by V_0 and force F by F_0 to give:

$$\bar{V} = \frac{e^{\gamma \bar{F}}\alpha - \beta}{\alpha - \beta} \qquad (2.24)$$

$$\text{where } \bar{F} = \frac{F}{F_0} \qquad (2.25)$$

$$\text{and } \bar{V} = \frac{V}{V_0} \qquad (2.26)$$

Equation (2.24) can be rewritten in terms of the single parameter γ (2.19) as:

$$\boxed{\bar{V} = \frac{e^{\gamma(1-\bar{F})} - 1}{e^{\gamma} - 1}} \qquad (2.27)$$

Figure 2(a) shows the variation of normalised mechanical velocity V_0 with normalised mechanical force $0 \le F_0 \le 1$ for three values of γ; the curve for $\gamma = 4.26$ corresponds to Figure 1 of Peskin et al. [4]. Figure 2(b) plots the same data as Figure 2(a) but in non-normalised form; this emphasises that both F_0 and V_0 increase with γ.

3 Power and efficiency

The *efficiency* of living systems in general, and actin-based transduction in particular, is of great interest [6, 7, 34]. As the bond graph approach directly exhibits energy flow, it allows the power and the efficiency of the actin filament to be directly investigated.

The mechanical power P_{mech} (flowing though the **TF**) is:

$$P_{mech} = FV \qquad (3.1)$$

This is, of course equal to that part of the chemical power flowing into the **TF** . The chemical power P_{chem} into the system from the two complexes represented by **Ce**:A and **Ce**:B is the product of the molar flow v and the chemical reaction potential Φ_c (2.22):

$$P_{chem} = v\Phi_c \qquad (3.2)$$
$$= vRT\gamma$$

Using the values from § 2.3, P_{mech} and P_{chem} are plotted in Figure 3(a). There are many concepts of efficiency, one possibility is to define efficiency η as:

$$\eta = \frac{P_{mech}}{P_{chem}} \qquad (3.3)$$

Using Equations (3.1), (3.2) and (2.23) is follows that efficiency η is proportional to normalised force:

$$\eta = \frac{FV}{v\Phi_c} = \frac{Fmv}{vmF_0} = \frac{F}{F_0} \qquad (3.4)$$

Figure 3: Power and Efficiency. (a) Using the values from Table 1 the normalised (by $P_0 = F_0V_0 = 2.03\,\text{aW}$) chemical power and mechanical power are plotted against normalised (by $F_0 = 6.75\,\text{pN}$) force. The chemical power reduces as force increases; the mechanical power (PV) is zero at two points: when the force $F = 0$ and the velocity $V = 0$ (corresponding to the stall force F_0). The difference between the two powers is dissipated in the reaction component **Re**:**r**. (b) The *efficiency* η (3.3) is plotted against normalised force $\bar{F} = F/F_0$; as indicated in Equation (3.4), the efficiency is equal to the normalised force. The efficiency is at its highest value when the system is in equilibrium: $V = 0$ and no energy is dissipated in the **Re** component. On the other hand, the highest power transfer occurs at $\eta \approx 20\%$; such is the tradeoff between power transfer and efficiency. (c) and (d). As (a) and (b) but plotted against normalised (by $V_0 = 0.30\,\mu\text{m}$) velocity.

This is plotted in Figure 3(b). As expected, the efficiency is at its highest value when the system is in equilibrium: $\bar{F} = 1$ and $\bar{V} = 0$ and no energy is dissipated in the **Re** component. On the other hand, the highest power transfer occurs at $\eta \approx 20\%$; such is the tradeoff between power transfer and efficiency.

4 Rigid filament not normal to surface

(a) Conceptual

(b) Bond graph model

Figure 4: Bond Graph model – rigid filament at angle $\theta \neq 0$ with normal to surface. (a) This is identical to Figure 1(a) except that the growing filament is at an angle of $\theta\mathrm{rad}$ to the membrane normal. As discussed in the text, the filament force F and velocity V are now different to the membrane force F_m and velocity V_m. (b) This is identical to Figure 1(b) except that the additional **TF** component **TF:c** is used to perform the geometric transformation between F,V and F_m,V_m imposed by the angle θ of Equations (4.1)& (4.2). It is assumed that the actin filament is in contact with a smooth surface and therefore produces no force parallel to the surface.

Mogilner and Oster [9] added two innovations to the model of Peskin et al. [4]: the actin rod impinges on the membrane at an angle θ to the normal ($\theta = 0$ corresponds to normal to the membrane, $\theta = \frac{\pi}{2}$ corresponds to parallel to the membrane); and the actin filament is flexible rather than rigid. The former is treated in this section, the latter in § 5. As discussed by Pollard [8] this work implies the existence of an optimal angle for maximum membrane velocity; this aspect is also examined in the bond graph context.

The rigid filament at an angle is shown in Figure 4(a) and the corresponding bond graph of Figure 4(b) is now derived. The bond graph approach models complex systems by incrementally adding more components. In the mechanical domain, the standard bond graph approach to modelling a coordinate rotation is to use an angle-modulated transformer **TF** component [15–17, 35]. With reference to Figure 4, the transformer **TF:c** has a modulus $\cos(\theta)$ and transforms F and V in rod coordinates to F_m and V_m in membrane coordinates. In particular:

$$V_m = V \cos(\theta) \tag{4.1}$$
$$F = F_m \cos(\theta) \tag{4.2}$$

Figure 5: Non-normal angle: $\theta \neq 0$: properties. (a) The normalised membrane velocity $\bar{V}_m = V_m/V_0$ is plotted against incidence angle θ for various values of the normalised membrane force $\bar{F}_m = \frac{F_m}{F_0}$. The optimal incidence angle for each value of force is marked. (b) The normalised membrane velocity $\bar{V}_m = V_m/V_0$ is plotted against normalised membrane force $\bar{F}_m = \frac{F_m}{F_0}$ for various values of the incidence angle θ. (c) The optimal incidence angle θ_{opt} is a function of membrane force F_m. (d) The effective stall force F_θ is a function of incidence angle θ.

Substituting Equations (4.1) and (4.2) into Equation (2.27) gives the formula giving normalised membrane velocity $\bar{V}_m$ in terms of normalised membrane force $\bar{F}_m$ with the parameters γ and θ:

$$\bar{V}_m = \cos\theta \frac{e^{\gamma(1-\bar{F}_m \cos\theta)} - 1}{e^\gamma - 1} \tag{4.3}$$

where

$$\bar{F}_m = \frac{F_m}{F_0} \tag{4.4}$$

$$\bar{V}_m = \frac{V_m}{V_0} \tag{4.5}$$

Using the same parameters as for Figure 2, and plotting normalised membrane velocity V_m against θ for a range of values of force F_m gives the curves of Figure 5(a); the vertical lines locate the optimal angle (giving the maximum membrane velocity) for each value of force. The optimum value of θ plotted against force is given in Figure 5(c). Note that it is possible to get forward membrane velocity for force greater than the zero stall force F_0 for $\theta > 0$. Indeed, the effective stall force is $F_0/\cos\theta$; this is plotted in Figure 5(d).

5 Flexible filament not normal to surface

If the actin filament is flexible, it will bend if a torque τ is applied; this could be analysed using standard beam theory [36]; but, as a first approximation, the flexible beam is replaced by a rigid beam of length L connected to the base by a rotational spring with compliance c radians/Nm as depicted in Figure 6(a). The base angle is θ_0 and the rigid filament angle is θ. The spring rotational angle ϵ is defined as

$$\theta = \epsilon + \theta_0 \tag{5.1}$$

Substituting for θ in Equation (4.3) gives:

$$\bar{V}_m = \cos(\theta_0 + \epsilon)\frac{e^{\gamma(1-\bar{F}_m \cos(\theta_0+\epsilon))} - 1}{e^\gamma - 1} \tag{5.2}$$

ϵ can be computed as follows. Using the equations for a linear simple spring with compliance c:

$$\epsilon = c\tau \tag{5.3}$$

$$\text{where the torque } \tau = F_m L \sin\theta \tag{5.4}$$

Combining Equations (5.1) – (5.4):

$$\epsilon = F_m cL \sin(\theta_0 + \epsilon) \tag{5.5}$$

$$= F_m cL(\sin\theta_0 \cos\epsilon + \cos\theta_0 \sin\epsilon) \tag{5.6}$$

(a) Conceptual

(b) Bond graph model

Figure 6: Bond Graph model – flexible filament. (a) This is identical to Figure 4(a) except that the growing filament is assumed to be flexible and the flexibility approximated by a spring with compliance σ $(\mathrm{N\,m\,rad^{-1}})$ at an angle of ϵ (rad) to the base normal. (b) This is identical to Figure 4(b) except that the **TF** component **TF**:c is replaced by a *modulated* transformer **MTF**:c modulated by the flexibility angle ϵ as generated by Equation (5.10) from the membrane force F_m.

(a) Angle-velocity (vary force)

(b) Force-velocity (vary angle)

(c) Optimum angle

(d) Force-velocity (vary flexibility)

Figure 7: Flexible filament. (a) Force/velocity diagram for various values of angle θ_0 and $\chi = 0.47$ (see § 5.1, Equation (5.17)). (b) Force/velocity diagram for various values of normalised compliance χ and $\theta_0 = 60°$. (c) The optimum (maximum velocity) values of θ are plotted for the flexible ($\chi = 0.47$) and rigid ($\chi = 0$) cases. (d) The velocity force diagram for the flexible filament; this should be compared to the rigid filament result of Figure 5(a).

Hence

$$\epsilon = \bar{F}_m \chi \left(\sin \theta_0 \cos \epsilon + \cos \theta_0 \sin \epsilon \right) \tag{5.7}$$

where the normalised compliance $\chi = cLF_0$ (5.8)

and $\bar{F}_m$ is given by (4.4). Following Mogilner and Oster [9, Appendix B], it assumed that $\epsilon \ll \pi/2$ and thus Equation (5.7) becomes:

$$\epsilon \approx \bar{F}_m \chi \left(\sin \theta_0 + \cos \theta_0 \epsilon \right) \tag{5.9}$$

hence:

$$\boxed{\epsilon \approx \frac{\chi \bar{F}_m \sin \theta_0}{1 - \chi \bar{F}_m \cos \theta_0}} \tag{5.10}$$

The corresponding bond graph is given in Figure 6(b). The modulated transformer **MTF** [35] allows the **TF** component to be modulated by ϵ (5.1) generated from the membrane force F_m using Equation (5.10).

The velocity-force curve is given by Equations (4.3), (5.1) and (5.10). Figure 7 summarises solutions to these equations for $\chi = 0.47$ (see § 5.1, Equation (5.17)) which can be compared to the rigid case of § 4 and Figure 5(d).

5.1 Computation of filament normalised compliance χ

The formula (5.10) contains the normalised compliance χ a a parameter. This section computes the value of χ corresponding to the parameters listed in Table 1.

Mogilner and Oster [9] use the *persistence length* λ as a measure of filament stiffness where:

$$\lambda = \frac{B_s}{k_B T} \tag{5.11}$$

where B_s is the filament *bending stiffness* ($B_s = EI$ in beam theory), k_B is the Boltzmann constant and T the absolute temperature. Modelling the filament as an Euler cantilever beam of length L with a force F_T at the tip, the deflection δ is [36, § 21, Chapter 3]:

$$\delta = \frac{F_T L^3}{3B_s} \tag{5.12}$$

Thus the deflection angle ϵ with respect to the root is:

$$\epsilon = \frac{\delta}{L} = \frac{F_T L^2}{3B_s} \tag{5.13}$$

Noting that the torque $\tau = F_T L$, the equivalent compliance c is given by:

$$c = \frac{\epsilon}{T} = \frac{L}{3B_s} \tag{5.14}$$

Using equation (5.11), the compliance c becomes:

$$c = \frac{1}{3k_BT}\frac{L}{\lambda} \tag{5.15}$$

hence, using equation (5.8):

$$\chi = \frac{L^2 F_0}{3k_BT\lambda} \tag{5.16}$$

Using the parameters from Table 1:

$$\chi = 0.47 \tag{5.17}$$

6   Experimental comparison.

Figure 8: Experimental comparison of the formulae (5.2) & (5.10) with experimental data presented by Li et al. [7] based on results of Bieling et al. [6]. The data is extracted from the additional information in elife-73145-fig1-data1-v2.xlsx. The value of $\theta_0 = 36°$ is taken from Li et al. [7]. Using the rigid model (§ 4, dashed line), the parameter γ (2.19) (the chemical driving force) is estimated. The estimated value is $\gamma = 3.07$. Using the flexible model (§ 5, firm line), the two parameters are estimated: γ and the normalised compliance χ (5.8). The estimated values are $\gamma = 3.15$ and $\chi = 0.29$.

Li et al. [7, Fig. 1D] provide experimental values of actin growth velocity for various values of growth stress obtained from branched actin network impinging on a cantilever. As they point out, force-feedback occurs in that the density of actin filaments increases with stress and this also appears in [7, Fig. 1D]. For the purposes of this paper, the behaviour of an average filament is obtained by normalising experimental pressure by the measured actin density[1].

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[1]Interpolation and extrapolation using a quadratic curve fit was used to create values of actin density for the values of experimental pressure in the velocity dataset.

This section fits the velocity-force curve for the flexible filament of § 5 given by Equations (5.2) & (5.10) to the normalised experimental data. There are 5 parameters: the angle θ_0, the chemical driving force encapsulated in γ (2.19), the normalised compliance χ (5.8), the stall force F_0 (2.18) and the unloaded velocity V_0 (2.15).

Li et al. [7] give the filament contact angle as $54°$. As the contact angle is relative to the surface (as opposed to the surface normal); this corresponds to $\theta_0 = 90 - 54 = 36°$. F_0 and V_0 were approximated by the maximum values of the normalised force and velocity respectively.

The remaining two parameters γ and χ were estimated by fitting the velocity-force curve for the flexible filament of § 5 given by Equations (5.2) & (5.10) to the normalised experimental data using the Python minimisation routine `scipy.optimize.minimize` using the default BFGS (Broyden–Fletcher–Goldfarb–Shanno) [37] method. This gives $\gamma = 3.15$ and $\chi = 0.29$; in comparison, the values of $\gamma = 4.26$ from Equation (2.21) and $\chi = 0.47$ from Equation (5.17) correspond to Table 1 and were used in Figure 7. The corresponding curve is plotted as the firm line in Figure 8 with the experimental points superimposed.

For comparison, the data was fitted with the rigid model ($\chi = 0$) by estimating the single parameter γ. This is shown as the dashed line in Figure 8. The deviation from experimental data is greater as normalised force increases; this corresponds to the the greater effect of flexibility at larger forces. The deviation would increase still further for larger values of membrane force F_m.

7   Conclusion

A novel bond graph approach to the modelling of actin chemomechanical transduction has been presented and has been shown to give an identical formula to the Brownian ratchet approach presented in the seminal paper of Peskin et al. [4]. The fact that two apparently disparate approaches give the same result seems to be due to the fact that bond graphs for chemical systems are, like the Brownian ratchet, built on thermodynamic principles [18, 19]. It is of historical interest to note that Oster was involved in establishing the foundations of not only the energy-based chemical bond graph [18, 19] but also the Brownian ratchet [4, 9].

The approach in this paper is closely related to previous work [23, 30] on the bond graph representation of chemoelectrical transduction where the **TF** component is used to link chemical and electrical domains using a modulus $m = z\mathbf{F}$ where z is the number of electrical charges and $\mathbf{F}$ ($\mathrm{C\,mol^{-1}}$) the Faraday constant. Indeed, the stall force formula (2.18) relating the chemical and mechanical domains is analogous to Nernst's equation relating the chemical and electrical domains.

Although, as emphasised by Peskin et al. [4] and Pollard [8], energy-based approaches "provide no mechanistic explanation of how the free energy of polymerization is actually transduced into directed mechanical force", the bond graph approach has the advantage that a model of an actin filament can be readily extended by adding additional components or modules. For example, it was shown in his paper that the basic model of actin chemomechanical transduction can be extended to non-normal incidence and flexible filaments by simply adding more components to the bond graph representation. More generally, the bond graph approach is inherently modular [21, 25, 26], and bond graph modules containing both basic bond graph components and

modules can be constructed in a hierarchical fashion. In this context, future work will investigate encapsulating the bond graph of actin chemomechanical transduction as a bond graph module and using the modular bond graph approach to incrementally build models of more complex systems including networks of actin filaments, adhesion, capping, moving membranes and actin recycling. The actin chemomechanical transduction is driven by ATP hydrolysis; this could be explicitly linked to bond graph models of metabolism. Using the approach of § 3, efficiency at the cellular level could then be investigated.

Understanding the flow of cellular energy is crucial to understanding life [34]; therefore energy-based bond graph modelling including the chemical, electrical and mechanical domains provides a route to this understanding.

8 Author Contributions

PJG prepared the numerical results and drafted the manuscript; PJG, MP and VR jointly developed the concepts, discussed the results and revised the manuscript. All authors approved the final version for publication.

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10 Data availability

The Python/Jupyter notebook used to generate the Figures is available at: `https://github.com/gawthrop/ActinRod24`.

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