

Frequency of elemental events of intracellular Ca^{2+} dynamics

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The dynamics of intracellular Ca^{2+} is driven by random events called Ca^{2+} puffs, in which Ca^{2+} is liberated from intracellular stores. We show that the emergence of Ca^{2+} puffs can be mapped to an escape process. The mean first passage times that correspond to the stochastic fraction of puff periods are computed from a novel master equation and two Fokker-Planck equations. Our results demonstrate that the mathematical modeling of Ca^{2+} puffs has to account for the discrete character of the Ca^{2+} release sites and does not permit a continuous description of the number of open channels.

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I. INTRODUCTION

Understanding the emergence of cellular processes from molecular interactions is one of the most fundamental quests in contemporary cell science. Since the number of reactions as well as the total number of molecules that participate in these reactions span orders of magnitude, no universal approach exists. That holds in particular when we consider the number of reacting molecules. On the one hand, there are processes that involve macroscopically large quantities, so that their dynamics is correctly described by deterministic equations [1,2]. On the other hand, recent experiments have revealed that some reactions affect only tens of molecules. Irvine *et al.* have reported that T cells react to the binding of even a single agonist [3]. Such a small number of interacting molecules demands a stochastic approach, because fluctuations cannot be neglected anymore as does deterministic modeling. In the MinCDE system for example, only a few thousand molecules are expressed [4]. It exhibits oscillations where the deterministic equations decay to a stationary fixed point. Research of recent years revealed several systems in which noise shapes the dynamics essentially and induces behavior that is not present without fluctuations (see [5,6] for examples from gene expression and signal transduction).

Another reason for the commitment of only a few molecules is the heterogeneity of cells. The number of reaction partners may exhibit strong spatial and temporal variations due to heterogeneously spatial distributions of molecules or large concentration gradients. These large gradients create a dynamic compartmentalization of the cell with largely different concentrations between the compartments. If only a small number of elements is in one of these dynamic compartments, fluctuations remain large, despite the fact that many copies of that element may be present in the cell. That is the case with intracellular Ca^{2+} dynamics (see later).

The dynamics of the Ca^{2+} concentration in the cytosol of a cell is determined to a large degree by release and uptake of Ca^{2+} by intracellular storage compartments, in particular the endoplasmic reticulum (ER). Release is controlled by inositol-1,4,5-trisphosphate (IP_3) receptor channels (IP_3R). They are arranged in clusters that comprise between 1 and 40 channels and that are randomly distributed on the membrane of the ER with distances between 1–7 μm [7,8]. IP_3R s have the important property that their open probability depends on

the Ca^{2+} concentration in the cytosol. The details of this dependency will be discussed in Sec. II. A moderate increase in the cytosolic concentration—i.e., on the outside of the storage compartment—increases the opening probability.

Single channels behave, of course, stochastically [9]. The most important fluctuations arise from the stochastic closing and opening of the Ca^{2+} channels [10]. They lead to random elemental release events called puffs. A puff is the spontaneous opening of channels of a single cluster. Experimental and theoretical studies have suggested that puffs play a pivotal role in intracellular Ca^{2+} dynamics [8,11]. These investigations put forward the idea that all global patterns like Ca^{2+} waves foot on Ca^{2+} puffs. To envisage the underlying mechanism, we start with a Ca^{2+} puff in a neighborhood of closed clusters. The Ca^{2+} released by the puff diffuses to adjacent clusters, where channels may open due to the Ca^{2+} dependent activation. If they do so, the liberated Ca^{2+} may induce neighboring clusters to open, too, and release spreads through the whole cell. That represents a single spike of an oscillation. However, there is no guarantee that a puff initiates a wave spreading throughout the cell, since activation is truly random, as we will show later.

The dynamic behavior of deterministic models using realistic concentration gradients provides further relevance of fluctuations for intracellular Ca^{2+} dynamics. Stochastic simulations and the bifurcation analysis of deterministic models have demonstrated that Ca^{2+} oscillations that agree with experimental findings vanish in the deterministic limit [10–12]. The reason is in the high Ca^{2+} concentration and large gradients that occur at an open cluster. They lead to a saturation of all control processes at the open cluster that regulate Ca^{2+} liberation in a deterministic model, so that no oscillations can occur [13,14]. Random fluctuations allow the cluster to escape from that saturation. The large gradients around an open cluster generate a dynamic compartment and keep the volume with high concentrations small. Consequently, only a small number of active elements experiences strongly stimulating concentrations. That precludes deterministic behavior in the dynamic compartment.

Given the vital part of Ca^{2+} puffs in intracellular Ca^{2+} dynamics and the importance of fluctuations, a stochastic description of a single cluster is the focus of the present work. We will apply it to the initiation of Ca^{2+} puffs, which represents the first step for any Ca^{2+} pattern. Our findings suggest that puff initiation can be mapped to an escape pro-

cess and that the mathematical description has to account for the integer number of open channels per cluster. A continuous model of the fraction of open channels that incorporates realistic fluxes does not permit Ca^{2+} puffs for parameter values that agree with experimental data. The mean first passage times can be represented as an infinite series of exponentials. However, already the first terms in the expansion yield excellent convergence. That hints at a Poissonian character of puff initiation. Although noise is intrinsically multiplicative for intracellular Ca^{2+} dynamics, we provide evidence that additive noise may serve as a reasonable approximation.

We will introduce a Ca^{2+} model for an IP_3 receptor channel cluster in the next section. It serves as input for a master equation in Sec. III, from which we will derive two Fokker-Planck equations in Sec. IV. Finally, we will employ these equations to characterize the initiation of Ca^{2+} puffs.

II. Ca^{2+} MODEL

The IP_3 receptor channel is a tetramer the subunits of which have binding sites for Ca^{2+} and IP_3 . We implement a model for a single subunit that is based on ideas of De Young's and Keizer's [15]. They assume a subunit to possess three binding sites: an activating Ca^{2+} binding site, an inhibiting Ca^{2+} binding site, and an activating IP_3 binding site. The occupation of the binding sites controls the state of the subunit. When IP_3 and Ca^{2+} are bound to their activating binding sites, a subunit is in the activated state. As soon as Ca^{2+} binds to the inhibiting binding site, a subunit is inhibited, independent of the state of the other binding sites. It can only be activated again upon dissociation of Ca^{2+} from the inhibiting binding site. Experiments have indicated that an IP_3 channel is conducting when at least three subunits are activated [16,17]. Random binding and unbinding of Ca^{2+} and IP_3 and therefore random state changes of the receptor are the source of stochasticity of intracellular Ca^{2+} dynamics.

The number of open IP_3 R channels determines the Ca^{2+} flux from the ER to the cytosol. Since the release channels are tightly packed within a cluster, a relation between the number of channels in the cluster and cluster size exists. Consequently, we can map the number of open channels to the size of a conducting area (or volume) equal to the area occupied by all open channels. A change in the number of releasing IP_3 R channels corresponds to a modulation of the conducting area of a cluster. This region is usually not connected. However, Swillens *et al.* showed that the spatial arrangement of IP_3 R channels does not influence the Ca^{2+} dynamics at an open cluster [7]. Therefore, we map the area of all conducting release channels to an area of the same size concentric to the cluster area. Let a denote the radius of this region, N the total number of channels per cluster, and n_o the number of open channels, then $a = a_0 \sqrt[3]{n_o/N}$. That reflects the above notion that the volume of the conducting sphere corresponds to the volume that is occupied by the fraction n_o/N of open channels. If $n_o = 0$ then $a = 0$, and a takes the maximal value a_0 if all N channels are open.

The deterministic dynamics of this cluster model has been investigated in [13,14]. In addition to IP_3 mediated Ca^{2+} liberation, we considered sarcoendoplasmic reticulum calcium

ATPase (SERCA) pumps, which transport Ca^{2+} from the cytosol to the ER, and a leak flux. The stationary Ca^{2+} concentration profile that results from these three fluxes is

$$c(r) = \left(A(a) \frac{\sinh(k_1 r)}{r} + e_1 \right) \Theta(a - r) + \left(B(a) \frac{\exp(-k_2 r)}{r} + e_2 \right) \Theta(r - a), \quad (1)$$

where

$$A(a) = \frac{l(k_2 a + 1)}{\cosh(k_1 a)k_1 + \sinh(k_1 a)k_2}, \quad (2a)$$

$$B(a) = \frac{l(\sinh(k_1 a) - \cosh(k_1 a)k_1 a)}{\exp(-k_2 a)[\cosh(k_1 a)k_1 + \sinh(k_1 a)k_2]}, \quad (2b)$$

with

$$l = \frac{-k_c k_p E}{(k_l + k_p + k_c)(k_l + k_p)}, \quad k_1 = \sqrt{\frac{k_l + k_p + k_c}{D}},$$

$$k_2 = \sqrt{\frac{k_l + k_p}{D}}, \quad (3)$$

and $e_1 := (k_l + k_c)E/(k_l + k_p + k_c)$, $e_2 := k_l E/(k_l + k_p)$. The constants k_l , k_p , and k_c denote the leak flux coefficient, the strength of the SERCA pumps, and the channel flux coefficient, respectively. The diffusion coefficient is given by D . E denotes the concentration of free Ca^{2+} in the ER.

Simulations have demonstrated that the Ca^{2+} concentration rapidly equilibrates upon a change in the number of open channels [12]. Hence, we will approximate the Ca^{2+} dynamics by its stationary value in the remainder of this work. The number of open channels n_o uniquely determines the Ca^{2+} concentration according to Eq. (1) and $a = a_0 \sqrt[3]{n_o/N}$. The focus of the two subsequent sections is the calculation of n_o .

III. MASTER EQUATION

The number of open channels n_o depends on the state of the subunits of the IP_3 Rs. A state of a subunit is determined by the occupation of its binding sites. The De Young Keizer (DK) model has three binding sites per subunit and hence eight subunit states. We reduce these eight states in two steps to three states. First, we eliminate the IP_3 dynamics adiabatically since IP_3 binding and unbinding are much faster than the Ca^{2+} dynamics in the framework of this model. The resulting four states are labeled by a binary pair ij , where the first index represents the Ca^{2+} activating binding site and the second the Ca^{2+} inhibiting binding site [11]. An index equals 1 when the binding site is occupied and 0 otherwise, e.g., 10 corresponds to the activatable state of a subunit. The second approximation uses the fact that we are interested in activation starting from a stationary state. Transitions among the inhibited states 11 and 01 have little impact on that activation process. Moreover, these states are rarely populated during puff initiation. Consequently we lump the two inhibited

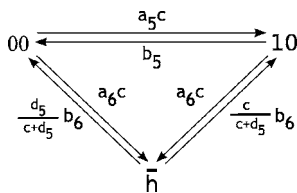


FIG. 1. Transition scheme for the three state model of the IP₃ receptor. $d_5 = b_5/a_5$ is the dissociation constant for Ca²⁺ activation, b_6 the dissociation rate of Ca²⁺ from the inhibiting site averaged over both IP₃ binding states. We denote the number of subunits in one of the three states by n_{10} , n_{00} , and $n_{\bar{h}}$.

states into one state \bar{h} . Figure 1 depicts the transition scheme for this three state model. The transition rates follow from [15,11].

Modeling the dynamics of an IP₃R on the basis of its subunits leads to various consequences for a cluster of N IP₃Rs. As long as every IP₃R is treated individually and subunits are assigned to individual channels—as has been done in stochastic simulations [11]—the state of the cluster is uniquely determined by the states of its subunits. However, an approach based on a population of subunits not grouped into individual channels is more suitable for the derivation of master equations and Fokker-Planck equations that we would like to use. That requires to determine the number of open

channels from the total number of activatable subunits in the subunit population. We assume that the activatable subunits are randomly scattered across the channels. The distribution of the n_{10} activatable subunits on the $4N$ subunits of a cluster decides upon the value of n_o and hence the Ca²⁺ concentration. We show in the Appendix that this distribution is sharply peaked around its mean value. Therefore, we set $n_o = \langle n_o \rangle = n_a$. n_a is defined in Eq. (A18).

The stochastic nature of Ca²⁺ release through IP₃Rs entails that the exact number of subunits in either of the three states 10, 00, or \bar{h} at a given time t , i.e., the triplet $[n_{10}(t), n_{\bar{h}}(t), n_{00}(t)]$, cannot be specified exactly anymore. On the contrary, only the probability $P(n_{10}, n_{\bar{h}}, n_{00}; t)$ to find a certain realization of $(n_{10}, n_{\bar{h}}, n_{00})$ at time t is accessible. Since the total number of subunits is fixed, the values of n_{10} and $n_{\bar{h}}$ suffice to specify the triplet $(n_{10}, n_{\bar{h}}, n_{00}; t)$, so that $P(n_{10}, n_{\bar{h}}, n_{00}; t) = P(n_{10}, n_{\bar{h}}; t)$.

The probability $P(n_{10}, n_{\bar{h}}; t)$ changes in the time interval $[t, t+dt]$ due to two opposing processes: Being in $(n_{10}, n_{\bar{h}})$ at time t , binding or unbinding of Ca²⁺ alters n_{10} or $n_{\bar{h}}$ during dt and hence reduces $P(n_{10}, n_{\bar{h}}; t)$. On the other hand, transitions from states as $(n_{10}+1, n_{\bar{h}})$ or $(n_{10}-1, n_{\bar{h}})$ into $(n_{10}, n_{\bar{h}})$ increases $P(n_{10}, n_{\bar{h}}; t)$. Taking all possible reactions according to Fig. 1 into account, the time evolution of $P(n_{10}, n_{\bar{h}}; t)$ is captured by the master equation [18],

$$\begin{aligned} \dot{P}(n_{10}, n_{\bar{h}}; t) = & - [n_{10}[b_5 + a_6c(n_{10})] + n_{\bar{h}}b_6]P(n_{10}, n_{\bar{h}}; t) - [hN - n_{10} - n_{\bar{h}}]a_5c(n_{10})P(n_{10}, n_{\bar{h}}; t) - [hN - n_{10} \\ & - n_{\bar{h}}]a_6c(n_{10})P(n_{10}, n_{\bar{h}}; t) + [n_{10} + 1]a_6c(n_{10} + 1)P(n_{10} + 1, n_{\bar{h}} - 1; t) + [hN - n_{\bar{h}} - n_{10} + 1]a_5c(n_{10} - 1)P(n_{10} \\ & - 1, n_{\bar{h}}; t) + [n_{10} + 1]b_5P(n_{10} + 1, n_{\bar{h}}; t) + \frac{b_6c(n_{10} - 1)[n_{\bar{h}} + 1]}{c(n_{10} - 1) + d_5}P(n_{10} - 1, n_{\bar{h}} + 1; t) + \frac{b_6d_5[n_{\bar{h}} + 1]}{c(n_{10}) + d_5}P(n_{10}, n_{\bar{h}} + 1; t) \\ & + [hN - n_{\bar{h}} - n_{10} + 1]a_6c(n_{10})P(n_{10}, n_{\bar{h}} - 1; t). \end{aligned} \quad (4)$$

For instance, being in $(n_{10}, n_{\bar{h}})$, the term proportional to a_6c in the first line denotes a transition from 10 to \bar{h} , so that the final state is $(n_{10}-1, n_{\bar{h}}+1)$. The Ca²⁺ concentration in the master equation depends on n_{10} , which is indicated by the notation $c(n_{10})$. The radius a in Eq. (1) follows from the number of activatable subunits as $a = a_0 \sqrt[3]{n_a/N}$ according to the preceding discussions. The adiabatic elimination of the IP₃ dynamics leads to noninteger values for the number of open channels. That demands a careful interpretation of the size of the conducting membrane patch, which was assumed to take only discrete values due to the discreteness of n_o . One approach is to truncate the rational values of n_o as $[n_o]_+$, where $[n_o]_+$ denotes the largest integer that is less or equal n_o . It entails $c = c_b$ as long as $n_o < 1$, where c_b denotes the base level of the Ca²⁺ concentration. This approach favors the closed configuration during puff initiation. In another approach we will keep the noninteger value of n_o and consider a as a quasicontinuous

function. We will discuss the effects of both approaches with respect to puff initiation.

Equation (4) is an accurate description of the stochastic dynamics represented by the scheme in Fig. 1. We will derive approximations like Fokker-Planck equations to calculate escape time characteristics from this master equation.

IV. FOKKER-PLANCK EQUATIONS

The discrete nature of master equations often impedes an analytic treatment. That holds in particular for master equations with nonlinearities or artificial boundary conditions. In these cases, several approximations have been put forward [18–22]. Despite the plethora of methods, there is still no consensus which approximation is best [23]. Each of them possesses advantages and drawbacks, so that the problem at hand finally decides which procedure to use. We will concentrate on van Kampen's Ω expansion and a method that is

similar to a Kramers-Moyal expansion. The latter keeps the nonlinearities of the master equation in the fluctuations, whereas the former approximates them in a linear fashion. Moreover, van Kampens's expansion is only valid when the macroscopic equation displays a single stable fixed point.

The Ω expansion requires a small parameter $1/\Omega$ in the master equation, which for our purposes is the inverse number of subunits, i.e. $\Omega=4N$. The systematic expansion of Eq. (4) in powers of Ω is based on the transformations $n_{10} = \Omega\phi(t) + \Omega^{1/2}\xi$ and $n_{\bar{h}} = \Omega\psi(t) + \Omega^{1/2}\eta$. They decompose the variables of the master equation into macroscopic parts (ϕ, ψ) and fluctuations (ξ, η). Inserting this ansatz into Eq. (4), the first nonvanishing order of Ω yields the macroscopic equations,

$$\frac{\partial\phi}{\partial t} = -\phi(a_5c + a_6c + b_5) + \psi\left(\frac{b_6c}{c+d_5} - a_5c\right) + a_5c, \quad (5a)$$

$$\frac{\partial\psi}{\partial t} = -(a_6c + b_6)\psi + a_6c, \quad (5b)$$

with $c = c(a_0\sqrt[3]{\phi_a})$ and $\phi_a := r^3\phi^3(4-3r\phi)$. r denotes the fraction of subunits in the state 10 that are activated: $r := I/(I+d_1)$. Equations (5) correspond to the rate equations that are associated with the transition scheme in Fig. 1, when the conservation condition $n_{10} + n_{\bar{h}} + n_{00} = 4N$ is applied. Note that ϕ_a is the continuous limit ($N \rightarrow \infty$) of Eq. (A18). Therefore, ϕ_a is the probability that at least three of the four subunits of an IP₃R are activated. The solutions of Eq. (5) represent the deterministic part of the above transformation of variables. They have the stationary values

$$\bar{\phi} = \frac{d_6c}{(c+d_5)(c+d_6)}, \quad \bar{\psi} = \frac{c}{c+d_6}, \quad (6)$$

which agree with results in [14]. $d_6 = d_2(I+d_1)/(I+d_3)$ is an effective dissociation constant. d_2 denotes the dissociation constant for Ca²⁺ inhibition when the IP₃ binding site is ligated, d_1 and d_3 represent the dissociation constants of IP₃ binding [15].

The next order in Ω determines the fluctuations through the probability $P(n_{10}, n_{\bar{h}}; t) = P(\Omega\phi + \Omega^{1/2}\xi, \Omega\psi + \Omega^{1/2}\eta; t) =: \Pi(\xi, \eta; t)$, according to

$$\begin{aligned} \frac{\partial\Pi}{\partial t} = & -\left(g_{11}\frac{\partial}{\partial\xi} + g_{21}\frac{\partial}{\partial\eta}\right)(\xi\Pi) - \left(g_{12}\frac{\partial}{\partial\xi} + g_{22}\frac{\partial}{\partial\eta}\right)(\eta\Pi) \\ & + \frac{1}{2}\left(h_{11}\frac{\partial^2}{\partial\xi^2} + 2h_{12}\frac{\partial^2}{\partial\eta\partial\xi} + h_{22}\frac{\partial^2}{\partial\eta^2}\right)\Pi. \end{aligned} \quad (7)$$

The matrices (g_{ij}) and (h_{ij}) with $h_{12} = h_{21}$ are defined as

$$\begin{aligned} g_{11} := & b_6d_5\psi c^1/(c+d_5)^2 - a_6(c + \phi c^1) \\ & - b_5 - a_5(c - (1 - \phi - \psi)c^1), \end{aligned} \quad (8a)$$

$$g_{21} := a_6c^1 - a_6\psi c^1, \quad (8b)$$

$$g_{12} := b_6c/(c+d_5) - a_5c, \quad (8c)$$

$$g_{22} := -(a_6c + b_6), \quad (8d)$$

and

$$h_{11} := a_5(1 - \psi - \phi)c + b_6\psi c/(c+d_5) + a_6\phi c + b_5\phi, \quad (9a)$$

$$h_{21} := -b_6\psi c/(c+d_5) - a_6\phi c, \quad (9b)$$

$$h_{22} := a_6(1 - \psi)c + b_6\psi, \quad (9c)$$

with

$$c^1 := \frac{dc}{da}(a_0\sqrt[3]{\phi_a})\frac{a_0}{3}\sqrt[3]{\phi_a^{-2}}[18r^4\phi^3 - 12r^3\phi^3 - 12r^3\phi^2]. \quad (10)$$

Equation (10) arises from inserting Eq. (A18) into $a = a_0\sqrt[3]{n_a}/N$ and then expanding $c(a)$ in powers of Ω . The matrix (g_{ij}) coincides with the matrix of the linearized macroscopic equations (5). The fluctuations enter through the matrix (h_{ij}) . The Hurwitz criterion [24] assures that this matrix is positive semidefinite, which means that Eq. (7) is a linear multivariate Fokker-Planck equation.

The linear treatment of the noise in Eq. (7) has cast some doubt on the validity of the Ω expansion. Therefore, a different class of Fokker-Planck equations have been proposed that keep the nonlinearities of the master equation. Kramers and Moyal have treated the shifts $n_{10} \pm 1, n_{\bar{h}} \pm 1$ of n_{10} and $n_{\bar{h}}$ in Eq. (4) by means of a Taylor expansion [19,20]. Following this procedure and defining the new variables $\phi := n_{10}/\Omega$ and $\psi := n_{\bar{h}}/\Omega$, we obtain a Fokker-Planck equation for the probability $p = p(\phi, \psi, t)$:

$$\begin{aligned} \frac{\partial p}{\partial t} = & \frac{\partial}{\partial\phi}\left(\phi a_6c + \phi b_5 - (1 - \psi - \phi)a_5c - \frac{b_6c}{c+d_5}\psi\right)p \\ & + \frac{\partial^2}{2\Omega\partial\phi^2}\left(\phi a_6c + \phi b_5 + (1 - \psi - \phi)a_5c + \frac{b_6c}{c+d_5}\psi\right)p \\ & + \frac{\partial}{\partial\psi}[b_6 - (1 - \psi)a_6c]p + \frac{\partial^2}{2\Omega\partial\psi^2}[b_6 + (1 - \psi)a_6c]p \\ & - \frac{\partial}{\Omega\partial\psi\partial\phi}\left(\phi a_6c + \frac{b_6c}{c+d_5}\right)p. \end{aligned} \quad (11)$$

The nonlinearities are introduced through $c = c(a_0\sqrt[3]{\phi_a})$ with ϕ_a defined as after Eq. (5). Equations (4), (7), and (11) constitute the starting point for a systematic study of puff frequencies. Given a configuration $(n_{10}^0, n_{\bar{h}}^0)$ at time $t=0$, they yield the probability for a configuration $(n_{10}^t, n_{\bar{h}}^t)$ at time $t > 0$. If we identify $(n_{10}^0, n_{\bar{h}}^0)$ with the resting state of a cluster and $(n_{10}^t, n_{\bar{h}}^t)$ with the first channel opening, such a transition in the configuration space gives the probability for a Ca²⁺ puff. Consequently, we interpret a puff as an escape process from the state $(n_{10}^0, n_{\bar{h}}^0)$ to the state $(n_{10}^t, n_{\bar{h}}^t)$. Although the previous equations allow the calculation of this escape probability, no general solutions are known for two dimensional escape processes (see [25] for a recent result). However, the

time scale separation between Ca^{2+} activation and Ca^{2+} inhibition leads to a reduction of the two dimensional equations to one dimension. Since the inhibiting processes are much slower than binding and unbinding of Ca^{2+} to the activating binding site, we assume that $n_{\bar{h}}$ remains unchanged during the initiation of a puff. That is identical to setting $n_{\bar{h}} = \text{const}$, and the master equation simplifies to

$$\begin{aligned} \dot{P}(n_{10}; t) = & -\frac{b_6 c(n_{10})}{c(n_{10}) + d_5} n_{\bar{h}} P(n_{10}; t) + \frac{b_6 c(n_{10} - 1)}{c(n_{10} - 1) + d_5} n_{\bar{h}} P(n_{10} \\ & - 1; t) - (4N - n_{10} - n_{\bar{h}}) a_5 c(n_{10}) P(n_{10}; t) + b_5 (n_{10} \\ & + 1) P(n_{10} + 1; t) - b_5 n_{10} P(n_{10}; t) + (4N - n_{10} - n_{\bar{h}} \\ & + 1) a_5 c(n_{10} - 1) P(n_{10} - 1; t) - a_6 c(n_{10}) n_{10} P(n_{10}; t) \\ & + a_6 (n_{10} + 1) c(n_{10} + 1) P(n_{10} + 1; t). \end{aligned} \quad (12)$$

For a later analysis, it is convenient to rewrite Eq. (12) in the form

$$\begin{aligned} \dot{P}(n_{10}; t) = & -(g_{n_{10}} + r_{n_{10}}) P(n_{10}; t) + g_{n_{10}-1} P(n_{10} - 1; t) \\ & + r_{n_{10}+1} P(n_{10} + 1; t), \end{aligned} \quad (13)$$

with

$$\begin{aligned} g_{n_{10}} = & \frac{b_6 c(n_{10})}{c(n_{10}) + d_5} [\Omega \bar{\psi}] + (4N - [\Omega \bar{\psi}] - n_{10}) a_5 c, \\ r_{n_{10}} = & b_5 n_{10} + a_6 c(n_{10}) n_{10}, \end{aligned} \quad (14)$$

and $\bar{\psi}$ defined as in Eq. (6). From Eq. (12), we could again derive Fokker-Planck equations in the same manner as before. However, a more direct approach for the one dimensional Ω expansion is setting η equal to zero in Eq. (7) due to $n_{\bar{h}} = \text{const}$. Keeping only the derivatives with respect to ϕ in Eq. (11) gives the nonlinear Fokker-Planck equation. Note that these one dimensional equations are only valid during the initiation phase of a puff, whereas Eqs. (4), (7), and (11) capture the full time evolution. Nevertheless, we will concentrate on Eq. (12) and the entailing Fokker-Planck equations in the remainder of this work, because they admit ana-

lytic solutions and provide far reaching insights into puff frequencies. The existence of analytic solutions is one of the most prominent features of van Kampen's expansion, so that we will treat the corresponding Fokker-Planck equation most generally in the next section.

V. ESCAPE TIMES

The initiation of a Ca^{2+} puff corresponds to an escape from the stationary state to the first channel opening. That requires the definition of the boundaries of the phase space area from which the escape occurs. Since we restrict the discussion to one dimension in phase space, the boundary consists of two points. We see from Eq. (12) that the lower boundary d is at $n_{10} = 0$ and that it is reflecting. That agrees with the interpretation of n_{10} as the number of activatable subunits, which is always positive. The value of the upper boundary b is chosen such that the number of open channels $n_o = 1$. The upper boundary corresponds to the escape site, so that the boundary condition is of absorbing type [26].

The time t to reach the absorbing boundary is a stochastic variable. It is described by the probability density $\rho(t)$, i.e., $\rho(t)dt$ is the probability that the stochastic process reaches b between t and $t+dt$. ρ is most conveniently computed from $G(x, t) = 1 - \int_0^t \rho(x, \tau) d\tau$, which represents the probability that $d \leq n_{10} < b$ at time t when it started at $x = n_{10}^0$ at $t=0$. The time evolution of G is governed by $\tilde{\mathcal{L}}$, which is the adjoint of the Fokker-Planck operator \mathcal{L} [26]. Up until now, no general solution has been obtained for arbitrary \mathcal{L} . Yet, an analytic expression exists for G in the case of a linear Fokker-Planck operator, e.g., van Kampen's Ω expansion. Since the following derivation always holds and is not restricted to the current problem, we introduce new constants v and w . They are given by $v = -g_{11}$ and $w = h_{11}$ defined as in Eqs. (8a) and (9a), respectively, in the present study. G obeys the linear backward Fokker-Planck equation [26],

$$\frac{\partial G(x, t)}{\partial t} = -vx \frac{\partial G(x, t)}{\partial x} + \frac{w}{2} \frac{\partial^2 G(x, t)}{\partial x^2}, \quad v, w > 0, \quad (15)$$

with initial and boundary conditions

$$G(x, 0) = \begin{cases} 1, & d \leq x \leq b, \\ 0, & \text{otherwise,} \end{cases} \quad \frac{\partial G(d, t)}{\partial x} = 0, \quad \forall t, \quad G(b, t) = 0, \quad \forall t. \quad (16)$$

The initial condition states that $d \leq x < b$ at $t=0$ with probability one. The reflecting boundary condition at $x=d$ in the adjoint Fokker-Planck equation is expressed by a no-flux boundary condition. Setting $G \equiv 0$ at the right boundary corresponds to an absorbing boundary. We solve Eq. (15) with the ansatz $G(x, t) = \exp(-\lambda t) u(x)$, $\lambda \geq 0$ so that it reduces to the ordinary differential equation,

$$\frac{d^2 u}{dx^2} - \frac{2vx}{w} \frac{du}{dx} + \frac{2\lambda}{w} u = 0. \quad (17)$$

Applying the transformation $z := x^2/4$, we find for $\bar{u}(z) := u(x)$,

$$z \frac{d^2 \bar{u}}{dz^2} + \left(\frac{1}{2} - \frac{4vz}{w} \right) \frac{d\bar{u}}{dz} + \frac{2\lambda}{w} \bar{u} = 0. \quad (18)$$

It equals Kummer's equation for $\bar{u}(\tilde{z}) := \bar{u}(z)$ with $\tilde{z} := 4vz/w$,

$$\tilde{z} \frac{d^2 \bar{u}}{d\tilde{z}^2} + \left(\frac{1}{2} - \tilde{z} \right) \frac{d\bar{u}}{d\tilde{z}} + \frac{\lambda}{2v} \bar{u} = 0. \quad (19)$$

Two independent solutions of Eq. (17) are [27]

$$u_1(x) := M\left(-\frac{\lambda}{2v}, \frac{1}{2}, \frac{vx^2}{w}\right), \quad u_2(x) := xM\left(\frac{1}{2} - \frac{\lambda}{2v}, \frac{3}{2}, \frac{vx^2}{w}\right). \quad (20)$$

M designates the confluent hypergeometric function,

$$M(a, b, x) := \sum_{k=0}^{\infty} \frac{(a)_k x^k}{(b)_k k!}, \quad (21)$$

where $(a)_0 := 1$ and $(a)_k := a(a+1) \dots (a+k-1)$. The boundary condition at $n_{10} = b$ entails that a solution of Eq. (17) is

$$v(x) := C_1 \left(u_1(x) - \frac{u_1(b)}{u_2(b)} u_2(x) \right) = u_1(x) - \frac{u_1(b)}{u_2(b)} u_2(x). \quad (22)$$

Without loss of generality, we set $C_1 = 1$ because it merely serves as a normalization. The second boundary condition fixes the still unknown eigenvalues λ . They constitute an infinite countable set $\{\lambda_n\}$ due to the finiteness of d and b . Therefore, the general solution of Eq. (15) can be expressed as

$$G(x, t) = \sum_{n=0}^{\infty} a_n \exp(-\lambda_n t) v_n(x). \quad (23)$$

The subscript of $v_n(x)$ indicates that Eq. (22) has to be evaluated at $\lambda = \lambda_n$ [see Eq. (20)]. The coefficients a_n are determined by the initial condition $G(x, 0)$, which results in

$$a_n = \frac{\int_d^b r(x) v_n(x) dx}{\int_d^b r(x) v_n^2(x) dx}, \quad (24)$$

$$r(x) := \frac{2}{w} \exp\left(-\frac{v}{w} x^2\right).$$

Here we used the orthogonality relation of the eigenfunctions $v_n(x)$:

$$\int_d^b v_n(x) v_m(x) r(x) dx = \delta_{m,n} \int_d^b v_n^2(x) r(x) dx. \quad (25)$$

The probability $\rho(x, t)$ that the absorbing state is reached between t and $t+dt$ is readily computed from $G(x, t)$ as $\rho = -\partial_t G(x, t)$. Note that ρ is already normalized due to the initial condition $G(x, 0)$. Hence, the mean first passage time $T(x)$ equals

$$T(x) := \langle t(x) \rangle = \int_0^{\infty} t \rho(x, t) dt$$

$$= - \int_0^{\infty} t \partial_t G(x, t) dt$$

$$= \sum_n \frac{a_n v_n(x)}{\lambda_n}. \quad (26)$$

Equation (26) includes an infinite number of eigenvalues. We found that the first three terms of the sum over n were sufficient to achieve results indistinguishable from the exact results of Eq. (30).

An alternative approach to the mean first passage time follows from the differential equation [26],

$$-vx \frac{dT(x)}{dx} + \frac{w}{2} \frac{d^2 T(x)}{dx^2} = -1, \quad (27)$$

with the solution

$$T(x) = \frac{2}{w} \int_x^b \frac{dy}{h(y)} \int_d^y h(z) dz, \quad h(x) := \exp\left(-\frac{v}{w}(x^2 - d^2)\right). \quad (28)$$

Performing the z integration, we find

$$T(x) = \sqrt{\frac{\pi}{vw}} \int_x^b dy \exp\left(\frac{v}{w} y^2\right) \operatorname{erf}\left(\sqrt{\frac{v}{w}} y\right) + \frac{\pi}{2v} \operatorname{erf}\left(\sqrt{\frac{v}{w}} d\right)$$

$$\times \left\{ \operatorname{erfi}\left(\sqrt{\frac{v}{w}} x\right) - \operatorname{erfi}\left(\sqrt{\frac{v}{w}} b\right) \right\}. \quad (29)$$

The functions $\operatorname{erf}(x)$ and $\operatorname{erfi}(x) = \operatorname{erf}(ix)/i$ denote the Gaussian error function and the imaginary Gaussian error function, respectively. The remaining integral can be solved by series expansion so that the final expression for the mean first passage time takes the form

$$T(x) = \frac{b^2}{w} F_{2;2}\left(1, 1; \frac{3}{2}, 2; \frac{v}{w} b^2\right) - \frac{x^2}{w} F_{2;2}\left(1, 1; \frac{3}{2}, 2; \frac{v}{w} x^2\right)$$

$$+ \frac{\pi}{2v} \operatorname{erf}\left(\sqrt{\frac{v}{w}} d\right) \left[\operatorname{erfi}\left(\sqrt{\frac{v}{w}} x\right) - \operatorname{erfi}\left(\sqrt{\frac{v}{w}} b\right) \right]. \quad (30)$$

We employed the generalized hypergeometric function,

$$F_{p;q}(a_1, \dots, a_p; b_1, \dots, b_q; x) = \sum_{l=0}^{\infty} \frac{(a_1)_l \dots (a_p)_l x^l}{(b_1)_l \dots (b_q)_l l!}, \quad (31)$$

and used the identity

$$\frac{j!}{2j+2} \sum_{l=0}^j \frac{(-1)^l}{(2l+1)(j-l)! l!} = \frac{1}{2} \binom{(1)_j (1)_j}{\left(\frac{3}{2}\right)_j (2)_j}. \quad (32)$$

We defer the proof to the Appendix. The reason for presenting two methods for evaluating the mean first passage time is based on their different scopes of applicability. If we were only interested in T , then Eq. (30) would be preferable be-

cause it requires less computation. However, we are limited to the first moment [26]. The advantage of the first approach is that we obtain any moment by one integration. Moreover, we have access to the time evolution of the escape process, which allows for a more detailed analysis.

The previous results could only be obtained analytically because the corresponding Fokker-Planck equation was linear. In the case of a nonlinear Fokker-Planck equation, all quantities have to be computed numerically. The mean first passage time is evaluated best from a generalization of Eq. (28). For $\mathcal{L} = -\partial_x A(x) + \partial_x^2 B(x)/2$ we find [26]

$$T(x) = 2 \int_x^b \frac{dy}{h(y)} \int_d^y \frac{h(z)}{B(z)} dz, \quad h(x) := \exp\left(\int_d^x \frac{2A(y)}{B(y)} dy\right). \quad (33)$$

The study of Fokker-Planck equations instead of master equations is often motivated by easier treatment. That holds in particular in higher dimensions, because a broader spectrum of tools is available for Fokker-Planck equations than for master equations [28] and even analytical calculations may be possible as in the case of Eq. (12). That constitutes one of the reasons for the derivations in Sec. IV. However, Fokker-Planck equations always represent approximations. The only way to test their quality is a comparison with results obtained from a master equation.

To this end, we consider a general one step process, to which class Eq. (12) belongs. We assume that this jump process starts at a site m at $t=0$. Being at site n the particle hops to the right with a rate g_n and to the left with a rate r_n , respectively. When it reaches the left boundary L , it is reflected. Then, the mean first passage time to arrive at a site $R > m$ reads [18] as

$$T_{R,m} = \sum_{i=m}^{R-1} \left(\frac{1}{g_i} + \sum_{l=L+1}^i \frac{r_l r_{l-1} \cdots r_l}{g_i g_{i-1} \cdots g_l g_{l-1}} \right). \quad (34)$$

That allows us to estimate the validity of the preceding approximations. The transition rates r_i and g_i follow from Eq. (14) for the current investigation.

VI. RESULTS

A. Mean first passage time

The calculation of the mean first passage times according to Eqs. (26), (33), and (34) necessitates a further specification of the boundaries. Since we consider a cluster with N channels, the upper boundary ϕ_b for the nonlinear Fokker-Planck equation is given by the solution of $(r\phi_b)^3(4-3r\phi_b) = 1/N$. The left hand side corresponds to the fraction of open channels, as discussed after Eq. (5). For the lower boundary, we have $\phi=0$. This value holds for the master equation as well. The upper boundary for the master equation is obtained by rounding off $\Omega\phi_b$ to its nearest integer value $[\Omega\phi_b]$. Before specifying the boundary conditions for van Kampen's expansion, we note that it describes the strength of the fluctuations ξ around the fixed point $\bar{\phi}$. The left boundary is imposed by $n_{10} > 0$, whereas the right boundary has to satisfy

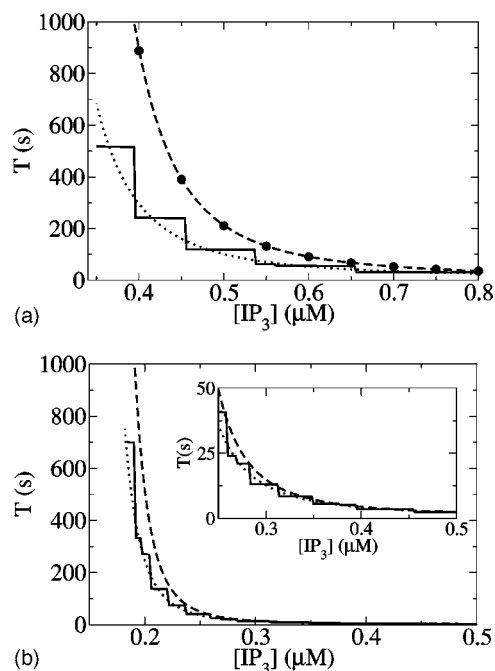


FIG. 2. Mean first passage time for $c_b=50$ nM (a) and $c_b=80$ nM (b) computed from the master equation (solid), the Ω expansion (dashed), and the nonlinear Fokker-Planck equation (dotted) for $d_1=0.13$ μM , $d_2=3$ μM , $d_3=0.9434$ μM , $d_4=0.4133$ μM , $d_5=0.24$ μM , $a_2=a_4=0.2(\mu\text{Ms})^{-1}$, $a_5=5(\mu\text{Ms})^{-1}$, and $N=25$. The dots in (a) represent the variance of the Ω expansion. The inset in (b) shows a blowup of the plot for large IP_3 concentration.

$\phi_b = \bar{\phi} + \Omega^{-1/2}\xi$. Consequently, the boundaries of ξ are $-\Omega^{1/2}\bar{\phi}$ and $(\phi_b - \bar{\phi})\Omega^{1/2}$, respectively, with $\bar{\phi}$ given by Eq. (6).

The mean first passage time depends strongly on the Ca^{2+} concentration [see, e.g., Eqs. (14) and (34)]. The results presented throughout Secs. VI A–VI C are calculated with a constant base level concentration. The number of open channels is an integer variable and there is no Ca^{2+} channel flux before the first channel opens. The Ca^{2+} concentration remains at a steady value until a Ca^{2+} puff occurs. That leads to $c^1 \equiv 0$ in Eq. (10) and to coefficients linear in ϕ and ψ in Eq. (11).

Figure 2 depicts the mean first passage time as a function of the IP_3 concentration for two different values of the basal Ca^{2+} concentration.

The master equation and the two Fokker-Planck equations exhibit an increase of the mean first passage time with decreasing IP_3 concentration. This increment diverges for lower values of the IP_3 concentration.

The nonlinear Fokker-Planck equation interpolates the master equation very efficiently. The results agree well with experimental findings for puff periods, although the mean first passage time only constitutes its mean stochastic fraction [8]. The discreteness of the master equation leads to discontinuities in the mean first passage time. The plateaus correspond to ranges of ϕ_b that are mapped to a single integer for the absorbing boundary of the master equation. Whenever that integer increases by 1, a jump occurs in the mean first passage time. van Kampen's expansion yields

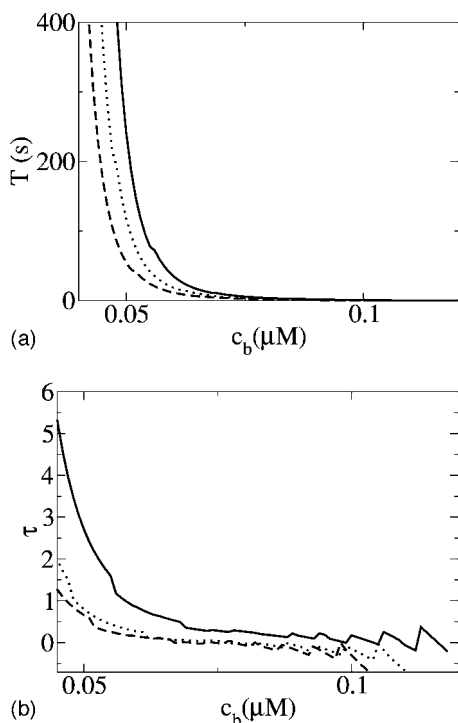


FIG. 3. Mean first passage time for the master equation (a) and the relative difference $\tau := (T_{vK} - T_{ME})/T_{ME}$ of the mean first passage time between van Kampen's method T_{vK} and the master equation T_{ME} (b) in dependence on the base level c_b for different values of the IP_3 concentration: $I=0.4 \mu\text{M}$ (solid), $0.5 \mu\text{M}$ (dotted), and $0.6 \mu\text{M}$ (dashed). Parameter values as in Fig. 2 and $a_5=1(\mu\text{Ms})^{-1}$.

good results for higher IP_3 concentrations, but overestimates the escape times otherwise (Fig. 2).

Figure 3 depicts the influence of the base level on the mean first passage time. The higher the basal concentration in this regime, the faster the first channel opens. van Kampen's expansion improves with increasing base level as a comparison between Fig. 2(a) and (b) and Fig. 3(b) shows. The zigzag behavior of the relative difference $\tau := (T_{vK} - T_{ME})/T_{ME}$ results from the discontinuities of T_{ME} ; see Fig. 2. Additionally, this quantifies the finding that the difference of the mean first passage time between the master equation and the Ω expansion diminishes with increasing IP_3 concentration.

B. Role of fluctuations

The most important difference between the nonlinear Fokker-Planck equation (11) and van Kampen's expansion (7) is in the diffusion term. It is constant in van Kampen's expansion—describing additive noise—and linear in ϕ and ψ in the nonlinear Fokker-Planck equation, thus describing multiplicative noise. As expected intuitively, the results in Fig. 2 show a better agreement between the nonlinear Fokker-Planck equation and the master equation than between van Kampen's expansion and the master equation. However, van Kampen's expansion approximates the master equation results rather well for high IP_3 and high base level of Ca^{2+} . That is quantified in Fig. 3. Consequently, additive

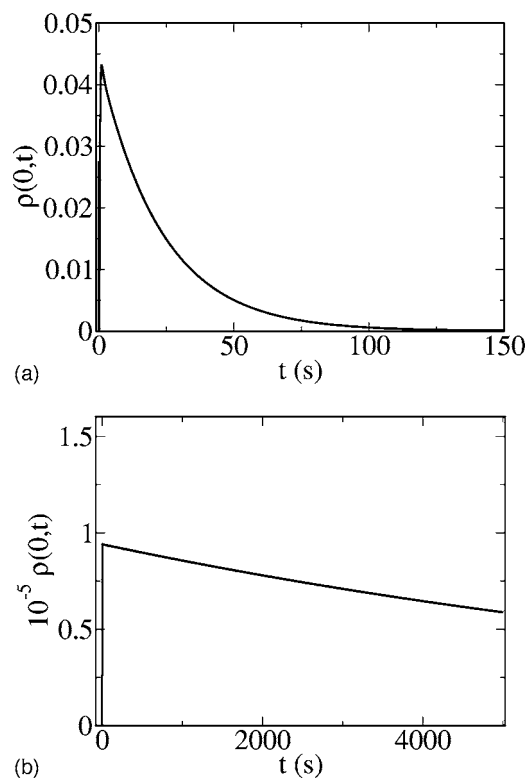


FIG. 4. Probability density $\rho(0,t)$ for van Kampen's expansion. Parameter values as in Fig. 2 and $I=0.5 \mu\text{M}$. (a): $c_b=0.06 \mu\text{M}$. (b): $c_b=0.04 \mu\text{M}$.

noise is probably a good approximation in these parameter areas.

C. Distribution of first passage time

van Kampen's expansion allows a direct computation of the probability density of the first passage time $\rho(0,t)$. $\rho(0,t)dt$ is the probability that the absorbing boundary is reached between t and $t+dt$. The starting point of the escape process in the Ω expansion is $\xi=0$. The IP_3 R cluster is exactly in the macroscopic state $\bar{\phi}$ at $t=0$, so that the noise vanishes at $t=0$. The results for ρ are depicted in Fig. 4. A convergence of the probability density according to Eq. (23) requires less than ten eigenvalues. The curves show the well known rising phase of ρ and the exponential decay. We find a maximal probability that shifts toward shorter times for higher IP_3 concentrations. The two graphs in Fig. 4 illustrate again the influence of the base level. Lowering c_b from 60 to 40 nM leads to an extreme broadening of the probability distribution and hence to an increase of the mean first passage time (see Fig. 3).

The probability density ρ permits an efficient computation of all moments of t for the escape process. Since the eigenvalues λ_n and the coefficients a_n are known, we immediately arrive at $\langle t^m \rangle = \sum_n m! a_n \lambda_n^{-m}$, in analogy to Eq. (26) due to $v_n(0)=1$ for all n . The dots in Fig. 2 depict the results for the variance. The first six eigenvalues suffice for an excellent convergence. That is a direct consequence of the spectrum of the backward Fokker-Planck operator in Eq. (15).

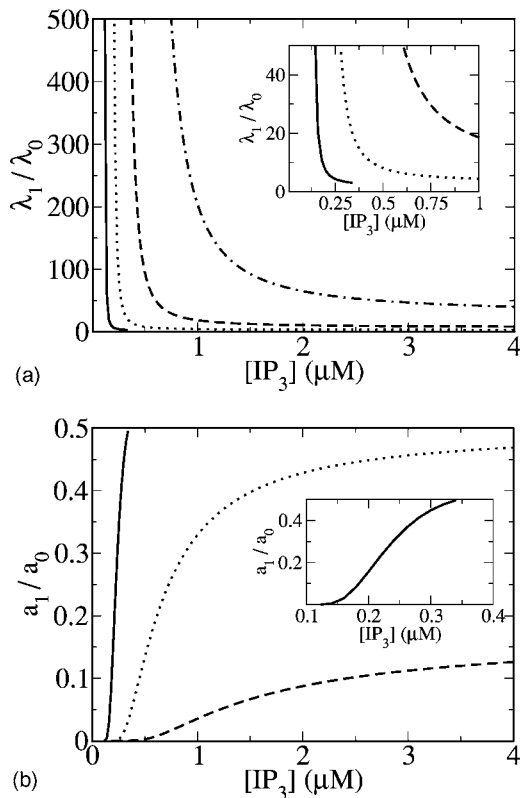


FIG. 5. Ratio of the first two eigenvalues (a) and the ratio of the first two expansion coefficients (b) of Eq. (23) in dependence on the IP_3 concentration. Parameter values as in Fig. 3 and $d_5 = 0.08234 \mu\text{M}$ (solid), $0.13234 \mu\text{M}$ (dotted), $0.183234 \mu\text{M}$ (dashed), and $0.23234 \mu\text{M}$ (chain-dotted). Insets show a blowup for small IP_3 concentrations.

Figure 5 shows the ratio of the first two eigenvalues λ_1/λ_0 . λ_1 is only a few times larger than λ_0 for large IP_3 and d_5 . However, the ratio increases with decreasing IP_3 concentrations and spans more than one order of magnitude for IP_3 concentrations smaller than $1 \mu\text{M}$. Hence, already the second term in the expansion (23) is considerably damped in the parameter range in which we are interested ($IP_3 < 1 \mu\text{M}$). Since the eigenvalues constitute a strictly increasing series, i.e. $\lambda_i < \lambda_j$ for $i < j$, the subsequent terms in the expansion decay even more rapidly. The prominent role of the first term is additionally supported by the expansion coefficients a_i . The ratio a_1/a_0 is depicted in Fig. 5(b). It decreases upon lowering the IP_3 concentration and tends to zero for very little concentrations. a_1 is much smaller than a_0 in parameter ranges where $\lambda_1/\lambda_0 \gg 1$ holds, i.e., where the second term of the series in Eq. (23) decays much faster than the first one. Consequently, higher terms only contribute marginally in this parameter regime. A detailed analysis of the spectrum and further implications will be provided in an upcoming report.

D. Continuous Ca^{2+} model

The results presented so far have been based on a discrete description of the number of open channels. The most important consequence is that the Ca^{2+} concentration remains constant as long as no channel opens. In the past, investiga-

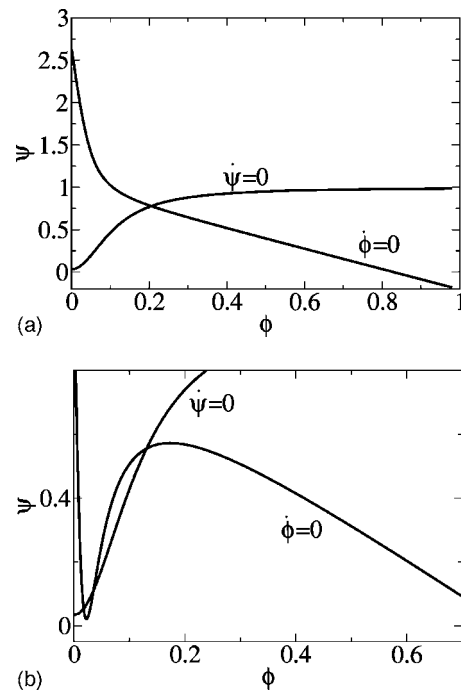


FIG. 6. Nullclines of Eq. (5). Parameter values as in Fig. 3 and $k_l = 0.002 \text{ s}^{-1}$, $k_c = 34500 \text{ s}^{-1}$, $k_p = 80 \text{ s}^{-1}$, $E = 750 \mu\text{M}$, $a_0 = 0.03 \mu\text{M}$, $D = 40 \mu\text{m}^2 \text{ s}^{-1}$, $d_5 = 0.1646 \mu\text{M}$, $I = 0.06 \mu\text{M}$ (a), $d_5 = 1.6468 \mu\text{M}$, $I = 0.053 \mu\text{M}$ (b).

tions on stochastic properties of IP_3R clusters often employed a continuous model of the ratio of open channels [29–31]. In these models, the Ca^{2+} concentration changes, even for fractions of open channels corresponding to less than one channel. Therefore, we have analyzed the impact of a continuously modulated number of open channels on the mean first passage time. The nullclines of the deterministic dynamics for such an ansatz with the same parameter values as before are displayed in Fig. 6(a). There is only one stationary state, which is linearly stable for all IP_3 concentrations. A prerequisite for a puff is that a sufficient number of subunits can be activated during the escape process from this fixed point. The value of ψ indicates that a large fraction of subunits is inhibited at already moderate IP_3 concentrations. It turns out that the remaining fraction of subunits is too low to induce a Ca^{2+} puff. The high degree of inhibition results from the large Ca^{2+} fluxes that occur at an open cluster [12]. These fluxes entail Ca^{2+} concentrations already in the μM range for sizes of the conducting area that are much smaller than that of a single channel. Since these concentrations exceeded the dissociation constants for inhibition, most of the subunits are inhibited. Lowering the IP_3 concentration does not invoke puffs, either. Although the fraction of inhibited subunits diminishes, the number of subunits that can be activated decreases as well.

We compensate for the elevated Ca^{2+} concentrations with an increase in the dissociation constant for Ca^{2+} activation, d_5 . The ensuing nullclines are depicted in Fig. 6(b). The left stationary state is linearly stable and corresponds to a low degree of inhibition. The motion of ϕ in phase space proceeds along an almost horizontal line through this fixed point

during puff initiation. These dynamics are bistable as the potential in Fig. 7 highlights. A Ca^{2+} puff parallels an escape process from the left well over the barrier to the first channel opening.

The time for such an escape process depends on the position of the absorbing boundary with respect to the barrier of the potential. Figure 7 shows the mean first passage time in dependence on the location ϕ of this boundary. ϕ varies from the value of the potential maximum (see the inset) to the value of the first channel opening ϕ_b (see Sec. VI A). The steep increase of T for small ϕ reflects the influence of the left well. As long as the absorbing boundary is close to the maximum of the potential, reentrance in the left well is possible. That becomes less dominant with increasing ϕ , so that the mean first passage time reaches the plateau. For the upper range of IP_3 concentrations in Fig. 7, the value of the plateau equals the mean first passage time. Consequently, the time scale of the puff is set by the properties of the left well. The strong increase of the mean first passage time for smaller IP_3 concentrations is due to two reasons. On the one hand, the left well of the bistable potential becomes broader and deeper with lower values of I . On the other hand, the absorbing boundary increases in a disproportionally high manner and moves higher on the right branch of the potential.

We exclude van Kampen's expansion in the above analysis, because its validity requires a single stationary state throughout the stochastic motion [18]. In contrast to a constant Ca^{2+} concentration, the nonlinear Fokker-Planck equation underestimates the results of the master equation. Nevertheless, the results in Fig. 7, which correspond to the stochastic fraction of the puff frequency, are in the same range as experimentally determined puff periods [8].

VII. DISCUSSION

We have derived a master equation and two Fokker-Planck equations for channel cluster behavior in IP_3 mediated Ca^{2+} dynamics. Among the different approaches to approximate a master equation by a Fokker-Planck equation we have chosen van Kampen's Ω expansion and an ansatz based on the Kramers-Moyal expansion. Master equations and corresponding Fokker-Planck equations for intracellular Ca^{2+} dynamics have been investigated in the past [29–31], but the study at hand is founded on different ideas. Most of the previous contributions employ the Li-Rinzel model [32] for the dynamics of a single subunit of an IP_3 receptor. It describes the time evolution of the fraction of subunits that are not inhibited yet, taking advantage of the time scale separation between IP_3 activation, Ca^{2+} activation, and Ca^{2+} inhibition. We have used a state scheme for one subunit that only eliminates the IP_3 dynamics adiabatically. It focuses on Ca^{2+} activation, which is the driving force behind puff initiation. Therefore, we consider Ca^{2+} activation as the fluctuating variable, whereas Ca^{2+} inhibition is the random variable in the Li-Rinzel model.

Our approach respects that time scale separation between Ca^{2+} activation and Ca^{2+} inhibition does not necessarily always hold.

The Ca^{2+} concentration plays a pivotal role in the initiation of Ca^{2+} puffs. On the one hand, it fixes the resting state

of a cluster, i.e., the starting point of the escape process. On the other hand, it determines the transition rates. The present work has demonstrated that the Ca^{2+} concentration needs to stay at the base level until the first channel opens. Theory provides Ca^{2+} puffs that are in agreement with experimental results [8] at physiological parameter values only if the Ca^{2+} concentration remains constant during the entire escape process. These findings underline the discrete character of IP_3R channels in a cluster [13,14].

We use a realistic value for the channel flux constant k_c in difference to earlier studies [29–31]. That value is based on detailed simulations [12] and leads to Ca^{2+} concentrations two to three orders of magnitude larger than base level at an open channel. That causes models with a continuous number of open channels to fail. The nonvanishing Ca^{2+} flux at fractions of open channels smaller than 1 resulted in highly elevated Ca^{2+} concentrations at a cluster due to the large flux density [12]. In turn, that induced a high degree of inhibition. Decreasing the IP_3 concentration reduced the level of inhibition, but the number of subunits that could be activated decreased, too. The lack of Ca^{2+} puffs was resolved by increasing the dissociation constant for Ca^{2+} activation d_5 . The ensuing mean first passage times again complied with experimental results, but at unphysiological values of d_5 . These results demonstrate that parameter values may decide upon the underlying mechanisms. The large Ca^{2+} fluxes demand a discrete modeling of the Ca^{2+} release channels. This discrete modeling is one of the aspects of this study, setting it apart from previous investigations of stochastic cluster dynamics [29,31].

At a constant Ca^{2+} concentration, the main difference between van Kampen's expansion and the nonlinear Fokker-Planck equation is in the character of fluctuations. They correspond to additive noise for the Ω expansion and to multiplicative noise in the latter approach. Although the noise is intrinsically multiplicative, van Kampen's expansion provides a reasonable approximation, which improves with an increasing base level and growing IP_3 concentration. It opens up the opportunity for further studies since the Ω expansion is the only method that yields analytic expressions for the probability density and all higher moments. That distinguishes it from the master equation and the nonlinear Fokker-Planck equation, for which only the first moment is directly accessible.

The dependencies of the mean first passage time on the Ca^{2+} concentration as well as on the IP_3 concentration comply with physiological findings. An increase of the basal Ca^{2+} concentration enhances the open probability of the IP_3R channel [16]. Consequently, the mean first passage time is to decrease with growing Ca^{2+} concentration. Our results fully agree with this activating role of Ca^{2+} (see Fig. 3). The same tendency was observed when we increased the IP_3 concentration, which agrees with the activating role of IP_3 .

The present study has provided a framework for the quantitative determination of Ca^{2+} puff frequencies. The mean first passage times correspond to the stochastic fraction of the interpuff interval, which is governed by the activation of the IP_3Rs . The second contribution to the interpuff interval is a deterministic part controlled by puff duration, inhibition, and recovery from it. Taking into account that Ca^{2+} puffs

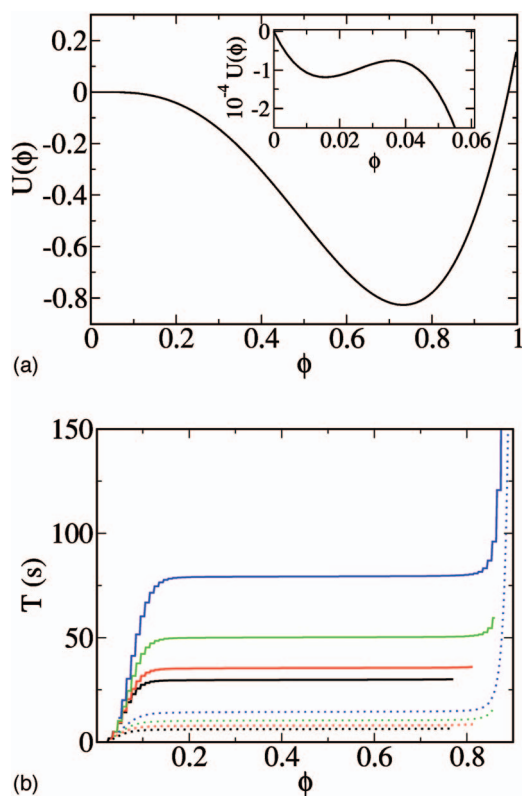


FIG. 7. (Color) (a): Potential $U(\phi)$ for $I=0.0483 \mu\text{M}$. The inset depicts an enlarged view for $\phi \approx 0$. Note the difference in scale for the axis. (b) (color): Mean first passage time computed from the master equation (solid lines) and the nonlinear Fokker-Planck equation (dotted lines) for $I=0.0553 \mu\text{M}$ (black), $0.0513 \mu\text{M}$ (red), $0.0473 \mu\text{M}$ (green), and $0.0433 \mu\text{M}$ (blue) in dependence on the position of the absorbing boundary ϕ . Parameter values as in Fig. 6(b).

represent the fundamental building blocks of global Ca^{2+} patterns, our calculations may serve as a starting point to compute periods of Ca^{2+} waves. Experiments and theoretical studies suggest that the initiation of Ca^{2+} waves occurs by a nucleation process. Therefore, knowledge of the frequency of Ca^{2+} puffs is the first step in the calculation of wave frequencies and leads to a deeper understanding of intracellular Ca^{2+} dynamics.

ACKNOWLEDGMENT

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APPENDIX A: COMBINATORICS FOR SUBUNITS

Measurements on the IP_3 receptor channel have revealed that a minimum number of subunits h_m needs to be activated for the channel to open [16]. A single IP_3R possesses a non-zero open probability only if at least h_m subunits are in the state 10. Activation in the cell occurs of course for a subunit already associated with a certain receptor. With our model, the number of open channels depends on the arrangement of n_{10} activatable subunits on the receptors. Here, we derive the

distribution of open channels resulting from such a random scattering of activatable subunits and its properties, whereas the mean was used earlier. To this aim we consider N receptors with h subunits each. Let $n_i, i=1, \dots, h$ denote the number of receptors with i activatable subunits, then the number of possible configurations for a given set $\{n_i\} := \{n_1, \dots, n_h\}$ that satisfies

$$n_0 + \dots + n_h = N, \quad n_1 + 2n_2 + \dots + hn_h = n_{10}, \quad (\text{A1})$$

is

$$M(\{n_i\}) := \frac{N!}{n_0! \dots n_h!} \binom{h}{0}^{n_0} \binom{h}{1}^{n_1} \dots \binom{h}{h}^{n_h}. \quad (\text{A2})$$

The fraction represents the number of permutations for the set $\{n_i\}$, whereas the binomial coefficients take into account the number of ways how to distribute i activatable subunits on a single receptor. The total number of configurations is given by

$$\Gamma := \sum_{\{n_i\}}^* M(\{n_i\}). \quad (\text{A3})$$

The asterisk indicates the summation with the restrictions of Eq. (A1). To evaluate Eq. (A3), we introduce a generating function,

$$f_l(z) := \sum_{\{n_i\}} M(\{n_i\}) z^l, \quad l = n_1 + \dots + hn_h. \quad (\text{A4})$$

The prime refers to the restriction $n_0 + \dots + n_h = N$. Therefore, the total number of configurations follows from the generating function as

$$\Gamma = \frac{1}{n_{10}!} \left. \frac{d^{n_{10}}}{dz^{n_{10}}} f_1(z) \right|_{z=0}. \quad (\text{A5})$$

Due to the identity

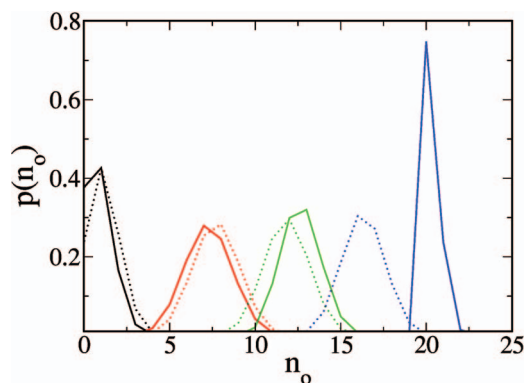


FIG. 8. (Color) Probability distribution $p(n_0)$ for $n_0 = n_3, h=3$ (solid) and $n_0 = n_3 + n_4, h=4$ (dotted) for $N=25$ and different n_{10} . Values of n_{10} are 25 (black), 50 (red), 60 (green), and 70 (blue).

$$f_1(z) = \sum_{n_1=0}^N \frac{N!}{n_0! \cdots n_h!} \binom{h}{0}^{n_0} \left[\binom{h}{1} z^1 \right]^{n_1} \cdots \left[\binom{h}{h} z^h \right]^{n_h} \\ = \left[\binom{h}{0} + \cdots + \binom{h}{h} z^h \right]^N = (1+z)^{hN} = \sum_{j=0}^{hN} \binom{hN}{j} z^j, \quad (\text{A6})$$

we finally arrive at $\Gamma = \binom{hN}{n_{10}}$, which complies with the combinatorics of choosing n_{10} subunits from a total of hN subunits. Consequently, the probability distribution of n_j for a fixed value of $j \in \{0, \dots, h\}$ is given by

$$p(n_j) = \frac{1}{\Gamma} \sum_{\substack{\{n_i\} \\ i \neq j}}^* \frac{N!}{n_0! \cdots n_h!} \binom{h}{0}^{n_0} \binom{h}{1}^{n_1} \cdots \binom{h}{h}^{n_h} \\ = \frac{1}{\Gamma} \binom{N}{n_j} \binom{h}{j}^{n_j} \sum_{\substack{\{n_i\} \\ i \neq j}}^* (N-n_j)! \prod_{\substack{l=0 \\ l \neq j}}^h \frac{1}{n_l!} \binom{h}{l}^{n_l}. \quad (\text{A7})$$

Equation (A7) is most conveniently computed as

$$p(n_j) = \frac{1}{\Gamma} \binom{N}{n_j} \binom{h}{j}^{n_j} \left. \frac{1}{n_{10}!} \frac{d^{n_{10}}}{dz^{n_{10}}} f_2(z) \right|_{z=0}, \quad (\text{A8})$$

where we used the generating function

$$f_2(z) := \sum_{\substack{\{n_i\} \\ i \neq j}} \tilde{N}! \prod_{\substack{l=0 \\ l \neq j}}^h \frac{1}{n_l!} \left[\binom{h}{l} z^l \right]^{n_l} \\ = \sum_{i=0}^{\tilde{N}} \sum_{l=0}^{hi} \binom{\tilde{N}}{i} \binom{hi}{l} \left[-\binom{h}{j} \right]^{\tilde{N}-i} z^{l+j(\tilde{N}-i)}. \quad (\text{A9})$$

Here, the prime denotes the restriction

$$n_0 + \cdots + n_{j-1} + n_{j+1} + \cdots + n_h = N - n_j =: \tilde{N}. \quad (\text{A10})$$

In the case $j=0$ the derivatives in Eq. (A8) can be performed explicitly, so that

$$p(n_0) = \frac{1}{\Gamma} \binom{N}{n_0} \sum_{j=0}^{\tilde{N}} \binom{\tilde{N}}{j} \binom{jh}{n_{10}} (-1)^{\tilde{N}-j}. \quad (\text{A11})$$

The previous analysis remains valid, when we interchange the number of activatable subunits n_{10} and the number of the remaining $Nh - n_{10}$ subunits. Such a transition corresponds to the exchange of balls and voids in classical combinatorics. In that picture, Eq. (A11) would represent the probability distribution of fully occupied receptors, i.e.,

$$p(n_h) = \frac{1}{\Gamma} \binom{N}{n_h} \sum_{j=0}^{\tilde{N}} \binom{\tilde{N}}{j} \binom{jh}{Nh - n_{10}} (-1)^{\tilde{N}-j}. \quad (\text{A12})$$

Equation (A12) arises from Eq. (A11) by substituting n_{10} by $Nh - n_{10}$ and n_0 by n_h . To gain further insight into the probability distributions we calculate the first two moments. For the average we start with

$$\langle n_j \rangle = \frac{1}{\Gamma} \sum_{\{n_i\}}^* n_j M(\{n_i\}), \quad (\text{A13})$$

because a closed expression for the probability distribution is only available for the two cases presented previously. Defining the corresponding generating function,

$$f_3(z) := \frac{1}{\Gamma} \sum_{\{n_i\}}^* n_j M(\{n_i\}) z^l, \quad l = n_1 + \cdots + hn_h, \quad (\text{A14})$$

we find

$$\langle n_j \rangle = \left. \frac{1}{n_{10}!} \frac{d^{n_{10}}}{dz^{n_{10}}} f_3(z) \right|_{z=0} = \frac{N}{\Gamma} \binom{h}{j} \binom{h(N-1)}{n_{10}-j}. \quad (\text{A15})$$

In the limit $N \rightarrow \infty$, $n_{10} \rightarrow \infty$ we recover the result from [33]. Analogously an evaluation of the second moments results in

$$\langle n_j n_k \rangle = \frac{N(N-1)}{\Gamma} \binom{h}{l} \binom{h}{k} \binom{h(N-2)}{n_{10}-l-k} \\ + \delta_{k,l} \frac{N}{\Gamma} \binom{h}{l} \binom{h(N-1)}{n_{10}-l}. \quad (\text{A16})$$

Applying these general expressions to IP₃R_s requires values for h , h_m , and N . The tetrameric structure of the receptor ensues $h=4$. However, previous results by different groups are based on $h=3$. We therefore compute the statistics for both cases. Experiments on a single channel have shown four conductance levels, each a multiple of 20 pS, with a predominance of opening to the third level [16,17]. Thus, we set $h_m=3$. The number of receptors in a cluster has not been measured yet. We employ $N=25$ following recent estimates by Swillens and Dupont [7].

The probability distributions $p(n_3+n_4)$ with $h=4$ and $p(n_3)$ with $h=3$ are depicted in Fig. 8. They both agree very well. This is also supported by their mean and variance, as shown in Fig. 9. In Fig.9(a) we also include the position of the maxima of the distributions indicated by dots. They closely follow the average. Due to the narrowness of the distributions demonstrated by the small variance as well as the accordance between the mean and the maximum we calculate the number of open channels n_c from the average for a given value of n_{10} ,

$$n_a^{(3)} = Nr^3 \frac{n_{10} n_{10} - 1}{3N} \frac{n_{10} - 2}{3N - 2}, \quad (\text{A17})$$

$$n_a^{(3,4)} = Nr^3 \frac{n_{10} n_{10} - 1}{4N} \frac{n_{10} - 2}{4N - 2} \left[\frac{n_{10} - 3}{4N - 3} (4 - 3r) \right. \\ \left. + 4 \left(1 - \frac{n_{10}}{4N} \right) \right]. \quad (\text{A18})$$

Here $r := l/(l+d_1)$ denotes the fraction of subunits in the activatable state 10 that are activated. The subscripts (3) and (3,4) indicate that we used $p(n_3)$, $h=3$ and $p(n_3+n_4)$, $h=4$ for averaging, respectively. Note that in the limit $N \rightarrow \infty$,

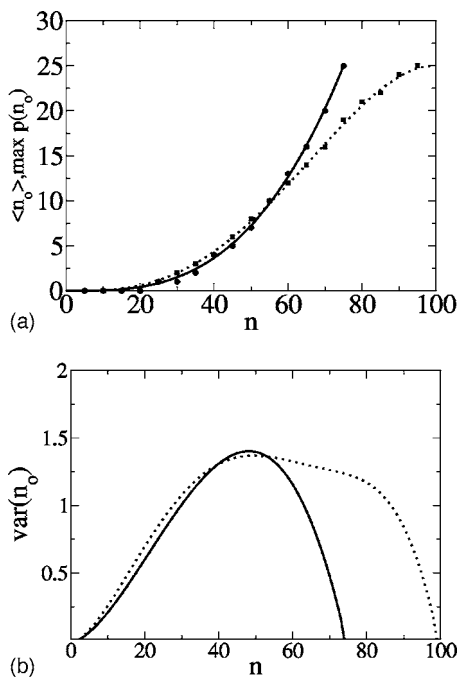


FIG. 9. Mean (a) and variance (b) of n_o for $n_o = n_3, h=3$ (solid) and $n_o = n_3 + n_4, h=4$ (dotted). (a) shows the position of $\max p(n_o)$ as dots and squares, respectively.

$n_{10} \rightarrow \infty$, Eqs. (A17) and (A18) reduce to the well known expressions of the deterministic description. All results in Sec. VI are based on Eq. (A18), which can be further simplified by approximating all denominators by $4N$ due to $4N \gg 1$.

APPENDIX B: PROOF OF EQ. (32)

In this section we deal with the proof of Eq. (32). It is based on the identity

$$\sum_{k=0}^j \binom{j}{k} \frac{(-1)^k}{2k+1} = \frac{2^{2j}(j!)^2}{(2j+1)!}, \tag{B1}$$

which we now prove. We transform the left hand side of Eq. (B1), according to

$$\begin{aligned} \sum_{k=0}^j \binom{j}{k} (-1)^k \int_0^1 t^{2k} dt &= \int_0^1 \sum_{k=0}^j \binom{j}{k} (-t^2)^k dt \\ &= \int_0^1 (1-t^2)^j dt. \end{aligned} \tag{B2}$$

It can be simplified with Euler's beta function $B(z, w)$. From its definition,

$$B(z, w) := \int_0^1 t^{z-1} (1-t)^{w-1} dt, \tag{B3}$$

it follows that

$$\int_a^b (t-a)^{z-1} (b-t)^{w-1} dt = (b-a)^{z+w-1} B(z, w). \tag{B4}$$

Hence we express the integral in Eq. (B2) through

$$\int_0^1 (1-t^2)^j dt = \frac{1}{2} \int_{-1}^1 (t+1)^j (1-t)^j dt = 2^{2j} B(j+1, j+1). \tag{B5}$$

According to [27], the beta function is related to the gamma function $\Gamma(z)$ via $B(z, w) = \Gamma(z)\Gamma(w)/\Gamma(z+w)$, so that we find

$$\sum_{l=0}^j \binom{j}{l} \frac{(-1)^l}{2l+1} = 2^{2j} \frac{\Gamma(j+1)^2}{\Gamma(2j+2)} = \frac{2^{2j}(j!)^2}{(2j+1)!}, \tag{B6}$$

due to $n! = \Gamma(n+1)$, which proves Eq. (B1). Expanding the right hand side yields

$$\begin{aligned} \frac{2^{2j}(j!)^2}{(2j+1)!} &= \frac{2 \cdot 1}{2} \cdot \frac{2}{3} \cdot \frac{2 \cdot 2}{4} \cdot \frac{2}{5} \cdot \frac{2 \cdot 3}{6} \cdots \frac{2 \cdot j}{2j} \cdot \frac{2}{2j+1} j! \\ &= \left(\frac{3}{2}\right)_j. \end{aligned} \tag{B7}$$

This proves Eq. (32) when we use $j! = (1)_j$.

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