Entropic Tension in Crowded Membranes

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Abstract

Unlike their model membrane counterparts, biological membranes are richly decorated with a heterogeneous assembly of membrane proteins. These proteins are so tightly packed that their excluded area interactions can alter the free energy landscape controlling the conformational transitions suffered by such proteins. For membrane channels, this effect can alter the critical membrane tension at which they undergo a transition from a closed to an open state, and therefore influence protein function in vivo. Despite their obvious importance, crowding phenomena in membranes are much less well studied than in the cytoplasm. Using statistical mechanics results for hard disk liquids, we show that crowding induces an entropic tension in the membrane, which influences transitions that alter the projected area and circumference of a membrane protein. As a specific case study in this effect, we consider the impact of crowding on the gating properties of bacterial mechanosensitive membrane channels, which are thought to confer osmoprotection when these cells are subjected to osmotic shock. We find that crowding can alter the gating energies by more than $2k_B T$ in physiological conditions, a substantial fraction of the total gating energies in some cases. Given the ubiquity of membrane crowding, the nonspecific nature of excluded volume interactions, and the fact that the function of many membrane proteins involve significant conformational changes, this specific case study highlights a general aspect in the function of membrane proteins.

Introduction

Cell membranes are packed full of proteins. The essence of various membrane inventories is that biological membranes are at least as much protein as they are lipid. Experiments on the occupancy of biological membranes by lipids and their protein partners provide a useful basis for making estimates of the possible consequences of membrane crowding. The presence of such high areal fractions of protein means that there is the possibility that the “crowding” effect can alter the free energies of different membrane protein conformations and the dynamics of the changes between these conformations as well. Indeed, over the last several decades, the importance of crowding effects in general has become a theme of increasing concern in physical biology [1–6].

The question of how the behavior of membrane proteins is altered by crowding effects has been explored much less thoroughly than their bulk counterparts [6–11]. As a concrete example of the way crowding might play out in membranes, we consider transmembrane proteins that have several conformations with different areal footprint. One particularly fascinating class of proteins of this variety are the mechanosensitive membrane channels. These proteins are thought to serve as safety valves for cells that are exposed to osmotic stress, opening up in response to increased membrane tension for the purpose of equilibrating the cells with their external environment [12–15].

To see how crowding might serve as an additional factor in the overall gating free energy balance for mechanosensitive channels, we consider the gating tension associated with the mechanosensitive channel of large conductance (MscL). Upon opening, at membrane tensions larger than $\approx 10^{-3}$ J/m$^2$, this channel undergoes a change in radius from roughly 2.4 nm to 3.5 nm [16–19]. As a result of this increased size, there is a reduction in the free area available for the surrounding membrane proteins resulting in an entropic driving force to keep the channel closed. The work presented here explores the relative importance of this effect compared to other contributions to the overall free energy budget for mechanosensitive channel gating.

In the remainder of this paper, we first examine various estimates of the degree of crowding in biological membranes. We then go on to explore the consequences of such gating for the free energy of the crowded proteins within the membrane, and the accompanying changes of the channel’s gating tension.

Results

The degree of crowding in membranes

As a prerequisite to characterizing the functional consequences of membrane crowding, we must first estimate the extent of crowding found in different types of membranes. There are various ways to arrive at numerical estimates of the extent of crowding of membrane proteins in biological membranes. One key measurable quantity that reflects the fraction of membrane area occupied by proteins is the protein to lipid mass ratio which typically falls in the range 1–2.5 [20–23]. Assuming that...
transmembrane (TM) domains make up about half of the membrane protein mass [24] and have roughly the same density as the lipids results in the estimate that 30–55% of the membrane area in the bilayer plane is occupied by proteins. Sowers and Hackenbrock [25] obtained electron microscopy images of mitochondrial inner membranes after application of a strong electric field that made all proteins drift to one end of the membrane surface, and found that the packed proteins in those images occupy 40–50% of the total area. Ryan et al. [26] fitted a statistical mechanics model of steric exclusion to the distribution of fluorescently labeled membrane proteins on rat basophilic leukemia cells subject to an electric field, and extracted an area coverage of 55–73%. Direct experimental estimates of the protein area fraction in red blood cell plasma membrane and synaptic vesicles have yielded area fractions of 20–25% [21,22]. In an extreme case, atomic force microscopy images of the photosynthetic membranes of *Rhodospirillum photometricum* cells [27] under various growth conditions show almost close-packed photosynthetic proteins arranged with nearly crystalline order. All of these examples tell the same fundamental story: membrane proteins are in very close proximity.

Another way of characterizing this crowding is by appealing to the number density which gives the number of membrane proteins per unit area of membrane. Aldea et al. [28] report that the five major outer membrane proteins (by mass) in *Salmonella typhimurium* have a total surface density of about 0.1/µm² in a wide range of growth conditions. Neidhardt et al. [29] (p. 41) quote lipoproteins as the most abundant protein (by number) in *Escherichia coli*, with ≈ 7 x 10⁵ copies in the outer membrane of a typical cell. Estimating the area of a typical *E. coli* to be 5 µm² [30], this gives a density of about 10⁷/µm²= 0.14/µm². Another way to estimate a protein density is to consider the fraction of the genome that codes for membrane proteins. In *E. coli*, about 1/3 of the 4200 genes encode membrane proteins, and the total number of proteins is about 3 x 10⁶ per cell [30]. If 1/3 of all proteins are evenly distributed in the two membranes, each membrane has about 500,000 proteins, or about 1 protein/µm². The areal and number densities estimated above are roughly consistent. If one assumes a footprint of 1.5 nm² per transmembrane helix [21–23], and 3 transmembrane helices per protein (see below), a number density of 0.1/µm² corresponds to an area fraction of 0.45.

There are other ways to think about the extent of membrane crowding, each with its own assumptions and merits, but regardless of these details the message will be the same. Biological membranes are crowded! For the purposes of this article, what these numbers tell us is that the mean spacing between proteins (estimated by evaluating 1/√C₉) is only slightly larger than the proteins themselves, so that a significant fraction of the membrane area is occupied by proteins.

Membrane proteins are not only abundant, they are also very heterogeneous, and vary significantly in size and shape [31]. Quantitative data on this heterogeneity is harder to come by, and we will therefore use the number of transmembrane helices (n_TMH) as an approximate indicator. Bioinformatic predictions of transmembrane regions [32] are routinely reported in surveys of proteins or putative protein-coding DNA regions [21,24,33], and range from one to several tens per protein subunit. Figure 1 illustrates two transmembrane helix distributions, based on a synaptic vesicle model [21] and a proteomics study of the outer membrane of the Gram-negative bacterium *Acinetobacter baumannii* [33], respectively. The latter is an average of three different techniques to estimate relative abundance, which differ significantly in specific cases, but lead to similar overall distributions (not shown). It is interesting to note the similarities in distributions in figure 1, both being dominated by proteins with a few TM helices, and spanning about one order of magnitude. However, there are several significant sources of uncertainty. For example, not all membrane proteins were detected [21,39], and we have not accounted for aggregation of protein subunits into larger complexes.

In our calculations below, we will model membrane proteins by circular disks, and will need to estimate \( \eta^2 = \text{Var}(R)/<R^2> \), where \( \text{Var}(R) = \langle(R - \langle R \rangle)^2 \rangle \) denotes the variance of \( R \). This quantity, which measures the variability of the projected protein area, enters into the more sophisticated treatments of the crowding effect discussed later in the paper. A useful approximation is \( \eta^2 \approx \eta_{TMH}^2 = \text{Var}(\sqrt{n_{TMH}})/<n_{TMH}> \), which comes from setting \( R^2 \) proportional to \( n_{TMH} \). Excluding proteins with no predicted transmembrane domains, the synaptic vesicle and *A. baumannii* outer membrane protein data sets in figure 1 give 0.25 and 0.14 for \( \eta_{TMH}^2 \), and 3.0 and 3.5 for the mean number of transmembrane helices, respectively. As we will see, these numbers indicate that size variability does not make a large quantitative contribution to the crowding effect, despite the quite broad distributions shown in figure 1.

**Figure 1.** Relative abundance of membrane protein subunits with different number of transmembrane (TM) helices. The histograms are based on data for synaptic vesicles [21], and the outer membrane (OM) of the Gram-negative bacterium *A. baumannii* [33]. Proteins with no predicted TM domains were excluded. doi:10.1371/journal.pcbi.1002431.g001
Crowding effects on gating

In light of estimated membrane protein crowding, our aim is to explore the implications of such crowding for channel gating. The total free energy change upon gating, $\Delta G_{\text{tot}} = G_{\text{open}} - G_{\text{closed}}$, can be thought of as arising from multiple contributions. In particular, we have

$$\Delta G_{\text{tot}} = \Delta G_{\text{protein}} + \Delta G_{\text{load}} + \Delta G_{\text{mem}} + \Delta G_{\text{crowd}},$$

where the first term reflects the free energy change associated with the protein degrees of freedom and their internal structural rearrangements, the second term refers to the potential energy of the loading device, and the third term characterizes the free energy of protein-lipid interactions, including the deformed membrane surrounding the protein that has been implicated as a key player in the gating of mechanosensitive channels [34–36].

The last term is the crowding-induced term. A membrane protein with a large cytosolic domain can potentially be crowded both by molecules in the cytoplasm, and by other membrane proteins. While the former effect has in fact been observed in the mechanosensitive channel MscS [37], it is the latter effect that forms the main substance of this paper.

The main conceptual point of the remainder of the paper can be stated simply as the idea that when the channel opens and changes its radius from “small” to “large”, there will be a free energy cost for the surrounding membrane proteins which we will refer to as crowders. In particular, these crowders will have their entropy reduced, which amounts to an effective pressure on the channel walls opposing its opening. To explore this claim, we will work in two distinct ensembles.

In the (mathematically) simpler case, we imagine a two-dimensional membrane “box” like that shown in figure 2A, such that the overall area is fixed. When the channel goes from the closed to the open state, there is a net reduction in the available area for the remaining crowders, which results in an entropic tension that favors the closed state. We make no reference to the elastic cost of squishing the lipids to access this state, since it can be shown that this energy is negligible in comparison with our main contribution of interest which is the entropic effect (see supporting text S1, Sec. 1).

The second scenario imagines a loading device that subjects the membrane to some fixed tension on its perimeter, much like the springs that hold a trampoline under its state of tension. It can be shown that in this case, when the channel goes from the closed to the open state, the areal strain, and hence the lipid area available to the crowders, do not change significantly [see supporting text S1, Sec. 2]. However, because of the change in the circumference of the protein, the exclusion annulus around the channel, indicated in figure 2, will be enlarged. Hence, there will still be an entropic tension which favors the closed state. In both cases, we make the implicit assumption that the number of lipids in the membrane does not change on the time scale of protein conformational changes.

To explore these two scenarios, we begin with the box of fixed area and use the simplest “ideal gas” physics to evaluate the change in entropy due to the loss of translational degrees of freedom when the channel goes from the closed to the open state. In particular, the translational entropy of one crowder can be computed as the logarithm of the area available to its center of mass,

$$S_{\text{crowd}}(R) = -k_B T \ln \left( \frac{L^2 - \pi (R + R_p)^2 - A_{\text{edge}}}{A_{\text{lattice}}} \right),$$

where $R_p$ is the radius of the crowder, $R$ is the radius of the channel, $A_{\text{edge}}$ is the band of thickness $R_p$ around the edge of the box from which the crowder center of mass is excluded (see figure 2A). The denominator $A_{\text{lattice}}$ refers to a discretization length scale used in a lattice model for the entropy [30]. Hence, the numerator is the effective area available to the crowder in the $L \times L$ membrane patch, recognizing that the minimal center-of-mass distance between the crowder and channel is $R_p + R$ (see figure 2).
This expression can be simplified by expanding in the small parameter $(\pi(R + R_0)^2 + A_{edge})/L^2$. If we exploit this simplification and add the contributions from $N$ crowded, the difference in the You energy between the open $(R = R_0)$ and closed $(R = R_c)$ states due to the crowding contributions can be written as $\Delta G_{\text{crowd}} = N \Delta G_{\text{crowd}} \approx c_c k_B T (\pi (R_c^2 - R_0^2) + 2\pi R_c (R_c - R_0))$, with $c_c = N/L^2$ being the crowder concentration. These two terms have simple and intuitive interpretations that are serviced by noting that we can rewrite the area and circumference change, respectively, as $\Delta A = \pi (R_c^2 - R_0^2)$ and $\Delta C = 2\pi (R_c - R_0)$. We can then divide the entropy-crowding tension into a surface and a line tension, and write

$$\Delta G_{\text{crowd}} \approx -\sigma_{\text{crowd}} \Delta A + \tau_{\text{crowd}} \Delta C,$$  \hspace{1cm} (3)

in the constant area ensemble. In our “ideal gas” approximation, the surface tension is $-\sigma_{\text{crowd}} = c_c k_B T$, the familiar ideal gas law. The line tension, $\tau_{\text{crowd}} = c_c R_c k_B T$, originates in the fact that the annulus of exclusion shown in figure 2 changes size upon gating. This contribution vanishes in the limit that the size of the crowders goes to zero. For both terms, we will need to appeal to our earlier estimates of protein areal concentrations to set the scale of the effect.

The treatment given above provides the simplest estimate of the crowding effect. However, as shown in figure 2C, things become more complicated in the high concentration limit. In particular, the amount of available area is much less than is suggested by the simple estimate above, where we made no reference to the way the crowders interact with each other. Neglecting these interactions understimates the crowding effects. For 50% protein area coverage, the more accurate computations described below give increased surface and line tension terms by a factor of four and two, respectively. Note that the entropic effect increases in a highly non-linear fashion with the crowder area fraction, effectively diverging as one reaches the closed-packing limit. The effect we describe can thus be potentially much larger than the already substantial estimates of $1 - 10 k_B T$ summarized below.

One way to think about this, illustrated in figure 2C, is in terms of exclusion zones around each crowder, analogous to the physics described by the van der Waals theory of gases. In the highly crowded regime, the theoretical difficulty is to compute the total size of the exclusion zones in a way that avoids double counting areas where multiple exclusion zones overlap. We use scaled-particle theory for mixtures of hard disks, an approximate equation of state that combines reasonable accuracy with analytical tractability [38–40], and has been widely applied to describe the effects of crowding [1,2,6–8,11,14]. The central results, for circular crowders, are presented in table 1, in terms of the concentration $c_c$, area fraction $\phi$, and, for non-uniform crowder size, relative size variance $\eta^2 = (\langle R_z^2 \rangle - \langle R_{\text{avg}}^2 \rangle)/\langle R_{\text{avg}}^2 \rangle$. Details of the derivations are presented in the Models section below. The crowding-induced changes in gating energy still take the form of Eqs. (3) and (4), with only the line tension contributing in the constant tension ensemble. The more exact scaled particle theory gives larger crowding tensions.

With these analytical results in hand, we now turn to the question of the actual magnitude of the crowding effect. To be concrete, we consider the case in which we have a membrane where the area is half lipids and half proteins (i.e. $\phi = 1/2$), big enough to make the ideal gas estimates questionable. We consider a radius change of a single channel from 2.4 to 3.5 nm (as is appropriate for MscL [18]), and a crowder radius of 1 nm. For simplicity, we also neglect size variability and set $\eta^2 = 0$ since using our estimate of $\eta^2$ in the range 0.15–0.25 would only give a $\sim 10\%$ correction. This leads to an estimated crowder density of $c_c = \phi / \pi R_c^2 \approx 0.16$ nm$^{-2}$. Using these numbers in the context of table 1, we get $-\sigma_{\text{crowd}} \approx 0.64 k_B T / \text{nm}^2$, and $\tau_{\text{crowd}} \approx 0.32 k_B T / \text{nm}$. This translates to crowding-induced changes in total gating energies of 15.2 and 2.2 $k_B T$, for the constant area and constant tension ensembles, respectively (see table 2). Even in the case of constant tension, most relevant to MscL, the crowding effect can have a sizable impact on channel gating.

**Models**

In this section, we review some basic results of scaled-particle theory, and derive the main results of the previous section and table 1. Motivations for some of the approximations we use, such as neglecting the area compression of the lipid bilayer, are given in the supporting text S1. See table 3 for a summary of the notation and symbols.

| Table 1. Entropic surface and line tensions induced by crowders. |
|---------------------------------------------------------------|
| $-\sigma_{\text{crowd}}/k_BT$ | $\tau_{\text{crowd}}/k_BT$ |
| Ideal gas | $c_c$ | $c_c R_c$ |
| SPT, uniform crowders | $c_c (1 - \phi)^2$ | $c_c R_c (1 - \phi)$ |
| SPT, non-uniform crowders | $c_c (1 - \phi)^2$ | $c_c (R_c)^2 (1 - \phi)$ |

Entropic surface and line tensions induced by crowders, estimated by an ideal gas calculation and scaled-particle theory (SPT, see Eqs. (20) and (27)). The results are derived for the case in which a single circular protein increases its radius from $R_0$ to $R_c$ in the presence of circular crowders with radius $R_c$. The non-uniform crowders case contains averages $< >$ over the crowder radius distribution, and this size variation $(\eta^2 = [R_c]/[R_{\text{avg}}^2] \geq 0)$ leads to a smaller surface tension effect compared to uniform crowders with the same mean size.

| Table 2. Estimated crowding effects on MscL gating. |
|---------------------------------------------------------------|
| $\Delta G_{\text{crowd}}$ | $\Delta \sigma_{\text{crowd}}$ | $\Delta \tau_{\text{crowd}}$ |
| IG | SPT | IG | SPT |
| constant area | constant tension |
| $4.3$ | $15.2$ | $1.1$ | $2.2$ | $k_BT$ |
| $0.21$ | $0.74$ | $0.05$ | $0.11$ | $k_BT / \text{nm}^2$ |

Different metrics for the effect of crowding on the gating behavior of a mechanosensitive channel. The first row shows the approximate changes in gating energies. The second row shows the corresponding increase in gating tension, $\Delta T_{\text{crowd}} = \Delta G_{\text{crowd}} / \Delta A$, which can be measured directly in patch-clamp experiments. For comparison, the typical gating tension for isolated MscL is 0.3–1.3 $k_BT / \text{nm}$. For MscL gating, the constant tension ensemble is the more appropriate model.

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Table 3. List of symbols.

| Symbol | Description |
|--------|-------------|
| $\tilde{\beta}_{TMH}$ | number of transmembrane helices |
| $\beta$ | inverse temperature scale |
| $\tilde{\alpha}$ | copy number vector |
| $n$ | total copy number |
| $\tilde{N}$ | copy number vector for the crowders |
| $N$ | total number of crowders |
| $A$ | area |
| $\sigma$ | surface tension (2D analog of negative pressure) |
| $c_A$ | areal number density |
| $R_i$ | in-plane radius of species $i$. In particular, $i = a,c$ for the open and closed channel conformations respectively. |
| $F$ | Helmholtz free energy (constant area ensemble) |
| $G$ | Gibbs free energy (constant tension ensemble) |
| $\Delta C$ | circumference change of channel |
| $\Delta A$ | area change of channel |
| $\langle R^m \rangle$ | $m$:th moment of the protein radius distribution |
| $\langle R \rangle$ | radius variance |
| $\langle R_p^m \rangle$ | $m$:th moment of the crowder radius distribution |
| $R_p$ | crowder radius, for the case of uniform crowder size |
| $\phi$ | area fraction of disks or proteins |
| $\rho$ | total area occupied by unstretched lipids |
| $\eta^3$ | relative protein radius variance |
| $A_A$ | unstretched equilibrium area in constant tension ensemble |

Basic results of scaled particle theory

In this section, we restate some basic results of scaled-particle theory that serve as the basis for our calculations. We will express the results in terms of area, temperature, and particle copy numbers, since this is what we will use as thermodynamic control variables, but also quote the results in terms of variables like concentration $c_A$ and area fraction $\phi$ (see table 3).

Scaled-particle theory has been generalized to heterogeneous mixtures of convex particles [42–44], but we restrict our attention to mixtures of circular disks [40]. We will simply quote the results we need, and refer to the literature for details on the derivations [6,38–40,45].

We start with the canonical partition function for a collection of hard disks with radii $R_j$ and copy numbers $n_j$, enclosed in an area $A$. The crowding effect we are interested in comes from the configurational entropy of the proteins, and we therefore omit velocities and internal degrees of freedom, and neglect boundary effects. The remaining configurational partition function depends on the many-particle interaction energy, $U(\{\tilde{x}\})$, for an ideal gas behavior due to the interaction energy $U$, which we take to be simple hard-disk repulsion. (By construction, $Q=1$ for an ideal gas, where $U=0$ for all configurations.)

For the computations below, we will break down configurational changes as removals and insertions of particles of different sizes, and also consider area changes as a result of changed particle size. We will therefore need the chemical potential and surface tension of the disk mixture that is our protein model.

Scaled particle theory offers a simple equation of state that relates the surface tension (2D analog of negative pressure) exerted by the disks to the area footprint, number density, and size variation of the disks. Rewriting for example Eq. (6.7) of ref. [40] in our notation, we get

$$Z(\tilde{\alpha},A,T) = Q(\tilde{\alpha},A,T) \prod_j \frac{A^{{n}_j}}{n_j!},$$

where $Q = \frac{1}{\tilde{\beta}} \int d^2\tilde{x} e^{-\beta U(\tilde{x})}$ describes the deviation from ideal gas behavior due to the interaction energy $U$, and write

$$\sigma_{SP} = \sqrt{\frac{\partial \ln Z}{\partial A}} \bigg|_{T,\tilde{\alpha}} = -\frac{c_A}{1 - c_A \pi \langle R^2 \rangle} - \frac{(\pi c_A \langle R \rangle)^2}{(1 - c_A \pi \langle R^2 \rangle)^2}.$$

(Note that we use the sign convention $\sigma dA$ for surface tension-area work, which is the opposite sign compared to the pressure-volume convention $-pdV$ used in the original derivations of scaled particle theory). After substituting $c_A = n/A$, this expression can be brought to the more compact form
\[ \frac{\sigma_{\text{SPT}}}{k_B T} = -n \frac{A - n \text{Var}[R]}{(A - n \langle R^2 \rangle)^2}, \quad (8) \]

where \([R] = \langle R^2 \rangle - \langle R \rangle^2\) is the disk radius variance. Note how the size variability decreases the (negative) pressure through the variance term.

This can again be rewritten in terms of concentration, area fraction, and relative size variability by combining the concentration \(c_A = n/A\) and the area fraction relation \(A - n \pi \langle R^2 \rangle = (1 - \phi) A\), and then the relation \(n \text{Var}[R]/A = c_A \pi \langle R^2 \rangle \phi^2 = \phi \sigma_{\text{SPT}}^2\), resulting in

\[ \frac{\sigma_{\text{SPT}}}{k_B T} = -c_A \frac{(1 - \phi) \pi \langle R^2 \rangle}{(1 - \phi)^2} = -c_A (1 - \phi \sigma_{\text{SPT}}^2). \quad (9) \]

When we consider area changes in the next section, the area integral of the surface tension at constant particle numbers will also come in handy, and we therefore integrate Eq. (8), and obtain

\[ \int \frac{\sigma_{\text{SPT}}}{k_B T} \, dA = -n \ln(A - n \pi \langle R^2 \rangle) + \frac{n^2 \pi \langle R^2 \rangle^2}{A - n \pi \langle R^2 \rangle}. \quad (10) \]

Finally, we will need the chemical potential. This is commonly divided into an ideal gas part plus a correction, called the excess chemical potential. Using \(\tilde{\mu} + \epsilon\) to denote the state with an added particle of species \(j\), the excess chemical potential is defined as the ratio

\[ \Delta \mu_j/k_B T = -\ln \frac{Q(\tilde{\mu} + \epsilon_j, T, A)}{Q(\tilde{\mu}, T, A)} \quad (11) \]

Using manipulations similar to those that lead to Eqs. (7) and (8), the scaled-particle theory approximation given by, e.g., refs. [6,40], can be rewritten in the form

\[ \Delta \mu_j/k_B T = -\ln \left(1 - \frac{n \pi \langle R^2 \rangle}{A}\right) + \frac{n(\pi R_j^2 + 2\pi \langle R_j \rangle)}{A - n \pi \langle R^2 \rangle} + \frac{(n \pi \langle R \rangle^2)}{A - n \pi \langle R^2 \rangle}, \quad (12) \]

where the averages should be computed with copy numbers \(\tilde{\mu}\), i.e., without the test particle present.

From the definition of \(\Delta \mu\) (Eq. (11)), one can see that \(e^{-\beta \Delta \mu} = Q(\tilde{\mu} + \epsilon_j, T, A)/Q(\tilde{\mu}, T, A)\) also has a probabilistic interpretation, namely as the probability that a test particle can be inserted somewhere in the fluid without overlapping with the other particles. This observation, which is exact, is in fact the starting point for one way to derive scale-particle theory (see e.g., [6,40]), by using a clever approximation to account for the overlapping exclusion zones in figure 2 C.

Finally, the chemical potential is given by the ratio of partition functions,

\[ \frac{\mu_j}{k_B T} = -\ln \frac{Z(\tilde{\mu} + \epsilon_j, T, A)}{Z(\tilde{\mu}, T, A)}, \quad (13) \]

If we substitute Eq. (6) and then Eq. (12), we get the scaled-particle approximation to the chemical potential, namely,

\[ \frac{\mu_j}{k_B T} = -\ln \frac{A_{j+1}^n}{(n_j + 1)!} \frac{A_{j+1}^n}{n_j + 1} + \Delta \mu_j = -\ln \left(\frac{A - n \pi \langle R^2 \rangle}{A - n \pi \langle R^2 \rangle}\right) + \frac{n(\pi R_j^2 + 2\pi \langle R_j \rangle)}{A - n \pi \langle R^2 \rangle} + \frac{(n \pi \langle R \rangle^2)}{A - n \pi \langle R^2 \rangle}. \quad (14) \]

**Gating transition in the constant area ensemble**

With the results of the previous section in hand, we are now in a position to derive our main results. Specifically, we will consider a situation with a single channel crowded by other proteins that do not change their configuration. We denote the copy number vector and total number of these crowders by \(\tilde{N}\) and \(N\) respectively. The state with a channel in state \(i\) (i.e., \(j = e, c\) for the open and closed state respectively) will then have the copy number vector \(\tilde{N} + \tilde{\epsilon}_i\).

In the results and discussion sections, we use \(G\) to denote a generic free energy. In the following derivations, we will be more precise, and use \(F\) and \(G\) for the free energy in the constant area and constant tension ensembles, respectively. In the thermodynamic limit, they are related by a Legendre transformation \(G(\tilde{\mu}, \sigma, T) = F(\tilde{\mu}, A, T) - \sigma A\), where \(\sigma\) is the surface tension.

When computing the gating energy changes, we expand in various small parameters. Specifically, we will consider the total area, or total lipid area, to be much larger than the area of a single protein of any species, but comparable to the total crowder footprint \(N \pi \langle R^2 \rangle\). This means that \(\pi R_j^2/A\) is a small parameter for all protein radii \(R_j\), but \(N \pi \langle R^2 \rangle/A(= \phi)\) is not small. In a typical E. coli cell, \(A = 5 \mu m^2\), which means that \(\pi R_j^2/A \sim 10^{-6}\) (for \(R_j = 3.5 \mu m\)). We will neglect such small terms.

To compute the free energy changes of a conformational change at constant total area, e.g., changing a particle from species \(i\) (say, a closed channel) to species \(j\) (say, an open channel), we subdivide the reaction into one insertion and one removal, by multiplying and dividing by the partition function of the intermediate state,

\[ \frac{\Delta F_{i \rightarrow j}}{k_B T} = -\ln \frac{Z(\tilde{N} + \tilde{\epsilon}_i, A, T)}{Z(\tilde{N} + \tilde{\epsilon}_i, A, T)} = \frac{-\ln Z(\tilde{N} + \tilde{\epsilon}_i, A, T)}{Z(\tilde{N} + \tilde{\epsilon}_i, A, T)} \quad (15) \]

Splitting the product of ratios, we can compare with Eq. (14) to identify the free energy change as the difference of chemical potentials for the two configurations,

\[ \frac{\Delta F_{i \rightarrow j}}{k_B T} = -\ln \frac{Z(\tilde{N} + \tilde{\epsilon}_i, A, T)}{Z(\tilde{N} + \tilde{\epsilon}_i, A, T)} + \ln \frac{Z(\tilde{N} + \tilde{\epsilon}_j, A, T)}{Z(\tilde{N} + \tilde{\epsilon}_j, A, T)}. \quad (16) \]

This means that we can use Eq. (14) with \(j = o, e\) to compute the entropic contribution to the free energy change. Using \(R_p\) to denote crowder radii, we get

\[ \frac{\Delta F}{k_B T} = \mu_j(\tilde{N}, A, T) - \mu_i(\tilde{N}, A, T) \quad (17) \]
To simplify, we first identify changes in area \( \Delta A = \pi(R_0^2 - R_c^2) \), and circumference \( \Delta C = 2\pi(R_0 - R_c) \), and set the coefficient of variation, \( g \), to zero.

Next, we use the same simplifications that lead to Eq. (9), and end up with

\[
\frac{\Delta F}{k_B T} = c_4\langle R_p \rangle \Delta C + c_4(1-\phi\eta^2)(1-\phi) \Delta A.
\]

The coefficients of \( \Delta C \) and \( \Delta A \) are the line and surface tensions tabulated on line three of table 1. The negative surface tension of the crowders (Eq. (9)) acts to oppose an increased radius of the protein, because increasing the protein footprint decreases the area available to the rest of the crowders. The quantities in these coefficients should be computed without the channel present (although computing them with the channel present would only make a small difference). The properties of the channel itself only enter through \( \Delta C \) and \( \Delta A \). We obtain the uniform crowders result (line 2 of table 1) as a special case, by replacing the mean radius by a single value, \( \langle R_p \rangle \rightarrow R_p \), and set the coefficient of variation, \( \eta \), to zero.

Next, we consider the constant tension ensemble, and show that we recover only the line tension effect, i.e., the \( \Delta C \) term, in that case.

### Gating transition in the constant tension ensemble

For the constant tension ensemble, the statistical mechanics recipe is to introduce an external tension \( \sigma \), i.e., an external loading device with energy \(-\sigma A \). We also include a term \( H_{\text{lipid}} \) for lipid elastic energy as a function of area, and integrate over all areas,

\[
\Xi(\vec{u},\sigma,T) = \int dA e^{-\beta(\sigma A - H_{\text{lipid}})} Z(\vec{u},A,T).
\]

As we show in the supporting text S1, real membranes are too stiff for changes and fluctuations in lipid area to give significant contributions to the gating energy of a single channel. This means that the above integral will be dominated by the area \( A_0 = A_0 + \pi \langle R^2 \rangle \), where \( A_0 \) is the total unstretched lipid area. To good approximation, we can therefore set \( e^{-\beta H_{\text{lipid}}} \approx \delta(A-A_0) \), and think of the lipids as having constant area and infinite stiffness. This makes it easy to evaluate the area integral,

\[
\Xi(\vec{u},\sigma,T) \approx Z(\vec{u},A_0,T) e^{\beta \sigma A_0},
\]

and we recover the free energy of the constant tension ensemble as the Legendre transformation of the free energy for the constant area ensemble,

\[
-G(\vec{u},\sigma,T) = -k_B T \ln \Xi(\vec{u},\sigma,T) \approx -k_B T \ln Z(\vec{u},A_0,T) - \sigma A_0.
\]

We now return to our test problem, and again denote the crowder copy numbers by \( N \), the presence of a channel in state \( i = a, c \) by \( N + \varepsilon_i \) etc. We can then divide the total free energy change into three contributions: removal of a closed channel at area \( A_{N+\varepsilon} \), an overall area change \( A_{N+\varepsilon} \rightarrow A_{N+\varepsilon} = A_{N+\varepsilon} + \Delta A \) with no channel present, and insertion of an open channel at area \( A_{N+\varepsilon} \):

\[
\Delta G = -\mu_c(N,A_{N+\varepsilon},T) \quad \text{Removing a closed channel.}
\]

\[
-\sigma A_0 + \int_{A_{N+\varepsilon}}^{A_{N+\varepsilon}} \sigma_{\text{SPT}}(\tilde{N}) dA + \mu_c(N,A_{N+\varepsilon},T) \quad \text{Inserting an open channel.}
\]

Area change with the channel absent.

Substituting Eq. (10) and Eq. (14), and assuming that the crowder background does not contain any other channels \( (N_a = N_c = 0) \), we get (after collecting terms)

\[
\frac{\Delta G}{k_B T} = -\ln(A_0 + \pi R_0^2) + \ln(A_0 + \pi R_c^2) - \frac{\sigma A_0}{k_B T} + N\pi(R_c^2 + 2\langle R_p \rangle R_c) - N\pi(R_0^2 + 2\langle R_p \rangle R_0) + \frac{N\pi(R_c^2)^2}{A_0 + \pi R_c^2} - N\pi(R_0^2)^2 \frac{A_0 + \pi R_0^2}{A_0 + \pi R_c^2}
\]

\[
N^2\pi^2\langle R_p \rangle^2 (R_0^2 - R_c^2) \frac{A_0^2}{(1 + \pi R_c^2/A_0)(1 + \pi R_c^2/A_0)}.
\]

Next, we Taylor expand in the small parameters \( \pi R_{a,c}/A_0 \), collect coefficients of \( \Delta C \) and \( \Delta A \) (most of which cancel), and end up with the following lowest order result:

\[
\frac{\Delta G}{k_B T} = \frac{N\langle R_p \rangle \Delta C}{A_0} - \frac{\sigma A_0}{k_B T} + \text{small terms,}
\]

Noting that \( N/A_0 = c_4/(1-\phi) \) and discarding the small terms, we finally get

\[
\frac{\Delta G}{k_B T} = \frac{\frac{c_4 N\langle R_p \rangle \Delta C}{A_0}}{1-\phi} - \frac{\sigma A_0}{k_B T}.
\]

Comparing with the constant area result of Eq. (20), we see that the contribution from the crowding surface tension has canceled, but that the coefficient of \( \Delta C \) is the same, namely the line tension in table 1. The extra term \( -\frac{\sigma A_0}{k_B T} \) reflects the work done by the loading device during the area change, and is independent of crowding conditions.

### Discussion

Membrane proteins in cellular membranes are crowded. Estimates based on data from a broad range of organisms and experimental techniques [20–28] indicate that membrane proteins occupy area fractions ranging from 20% to well over 50% in different cell membranes. Crowding induces an entropic tension in the membrane, which favors membrane protein conformations...
with smaller areal footprint and circumference. This effect can be understood qualitatively using simple free area arguments, but quantitatively meaningful estimates require more sophisticated theories. We have used scaled-particle theory for hard disk mixtures [6–8, 38–40, 45] to compute the crowding induced line and surface tensions (see Models, Eqs. (20) and (27), and table 1). As a case study, we apply these results to estimate the influence of crowding on the gating tension of the bacterial mechanosensitive channel MscL. This channel is thought to act as a safety valve for cells under osmotic stress, opening up in response to high membrane tension in order to avoid membrane rupture [12–15].

There are different ways to quantify the influence of crowding on gating behavior (see table 2). One way to present the significance of our results is by appealing directly to the curves that provide the probability of channel opening as a function of the driving force. For the case of a “two-state” channel, which transitions back and forth between distinct closed and open states, the open probability is 

\[ p_{\text{open}} = \frac{1}{2} \left( 1 + \exp(\Delta G_{\text{tot}}/k_BT) \right)^{-1} \]

where \( \Delta G_{\text{tot}} \) is the energy difference between the closed and open states, and depends upon the driving force (such as tension, voltage or ligand concentration). In our case, the driving force is the tension, and we can rewrite it in the simpler form \( \Delta G_{\text{tot}} = \Delta G_0 - \Delta A + \Delta G_{\text{crowd}} \). The first term corresponds to all contributions of Eq. (1) that do not depend explicitly on crowding or applied tension. We can rewrite it in the simpler form \( \Delta G_0 = \sigma_0 \Delta A \). Figure 3 shows the gating probability \( p_{\text{open}} \) as a function of both for a single isolated channel and for the case in which crowders are present.

An alternative way to decide if the effect is big or small, is to compare it to some reference energy (or tension). The first relevant energy scale for comparison is the thermal energy \( k_BT \), the energy scale in the Boltzmann weight \( \exp(-\Delta G/k_BT) \) in the open probability above. Our numerical examples in table 2 all change the gating free energy by \( \geq 1 \, k_BT \). A second relevant energy scale is that associated with the gating of various mechanosensitive channels. The gating properties of channels such as MscL have been measured using several different species of lipid molecules in the surrounding membrane. The outcome of these elegant experiments is that the gating energies have typical values of 5–20 \( k_BT \) [17,46] and corresponding gating tensions in the range of 0.3–1.3 \( k_BT/nm^2 \). In the presence of spontaneous curvature inducing lipids, these energies and tensions are even smaller (or even negative, meaning that the channel opens spontaneously without any applied tension) [17]. The change in gating tension due to crowding is \( \Delta G_{\text{crowd}} = \Delta G_{\text{crowd}}/\Delta A \), and we get numbers in the range \( 0.05 - 0.7 \, k_BT/nm^2 \).

The entropic cost of channel opening in a crowded solution of membrane proteins has so far been discussed only with reference to hard core repulsion between proteins. It is however well known that membrane-mediated interactions may emerge from the overlap of the membrane deformations surrounding neighboring proteins, such as those arising from a thickness mismatch between the hydrophobic protein core and the membrane average thickness [34, 35], or a non-cylindrical shape of the transmembrane region [34, 35, 47]. Beside the hydrophobic mismatch itself, the strength, and even the sign of such interactions depend on many factors, including membrane stiffness to bending and stretching, and the monolayer’s spontaneous curvature. The range of these interactions is comparable to the protein size itself, and hence could be expected to influence the effect of crowding on the gating energy significantly.

The rich and interesting many body effects that can emerge from local membrane deformations are outside the scope of this paper. However, our calculations offers some qualitative insight into the sensitivity of the crowding effect to structural features of the involved proteins, which also includes some effects of hydrophobic mismatch. In our hard disk calculations under constant tension, the entropic surface tension cancels from the gating energy contribution (between Eqs. (3) and (4), and in Eq. (25)), when the increase in channel area is balanced by an increased total area. This cancellation reflects an underlying assumption in the disk model of membrane proteins, which effectively models membrane proteins as cylinders (figure 4A), from which lipids and other proteins experience the same area exclusion.

Real membrane proteins, however, can have complex shapes that violate this assumption [31], for example due to large domains outside the bilayer that do not directly affect the local ordering of the lipids, but provide additional steric interaction with other membrane proteins, as sketched in figure 4B. Many membrane bound receptors that bind bulky ligands near the membrane surface [48–51], yield complexes with a similar shape. We would expect significant crowding effects on both the binding kinetics and the stability of the complex for these systems, similar to what has been seen for surface adsorption [9–11,41]. There are also examples of membrane proteins whose bulky cytoplasmic domains undergo substantial conformational changes, such as the mechan-

![Figure 3. The effect of crowding on the open probability as a function of applied tension.](image)

**Figure 3. The effect of crowding on the open probability as a function of applied tension.** The graphs illustrate the ideal gas (\( \Delta G_{\text{crowd}} = 1.1 \, k_BT \)) and scaled-particle theory (SPT, \( \Delta G_{\text{crowd}} = 2.2 \, k_BT \)) results of table 2, using the constant tension ensemble as is appropriate for MscL. All non-crowding contributions to the gating free energy are lumped together in the gating tension \( \sigma_0 \). doi:10.1371/journal.pcbi.1002431.g003

![Figure 4. Mechanisms for different excluded area for proteins and lipids.](image)

**Figure 4. Mechanisms for different excluded area for proteins and lipids.** One difference between the hard disk model of membrane proteins (A), and more complex protein structures (B, C) might be thought of in terms of different effective radii \( R \) and \( \tilde{R} \) for steric exclusion of surrounding proteins and lipids respectively. A protein (red) with a large domain outside of the bilayer (B) might exclude surrounding proteins, but not lipids, from approaching the transmembrane domain (dark red). Similarly, proteins with different hydrophobic thickness than the surrounding bilayer (C) generate a local zone of deformed lipid bilayer (gray) that effectively excludes other well-matched proteins. Horizontal lines indicate the surrounding lipid bilayer. doi:10.1371/journal.pcbi.1002431.g004
The present analysis has as its key outcome the hypothesis that under sufficiently crowded conditions, membrane proteins can influence each others conformational changes through an entropic tension. Though we explored the consequences of that idea for one particular channel, given the great diversity of membrane proteins and the high degree of crowding in many membrane types, we expect that such effects could be common.

Supporting Information

Text S1 Additional calculations to justify some assumptions mentioned in the main text. (TEX)

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Author Contributions

Conceived and designed the experiments: ML PS RP. Performed the experiments: ML PS RP. Analyzed the data: ML. Wrote the paper: ML PS RP.

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