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Simplified proton exchange membrane fuel cells for space and terrestrial applications
Water uptake profile in a model ion-exchange membrane: Conditions for water-rich channels

Daniel C. Herbst,1,a) Thomas A. Witten,1 Tsung-Han Tsai,2 E. Bryan Coughlin,2 Ashley M. Maes,3 and Andrew M. Herring3
1James Franck Institute and Department of Physics, University of Chicago, Chicago, Illinois 60637, USA
2Polymer Science and Engineering Department, University of Massachusetts, Amherst, Massachusetts 01003, USA
3Chemical and Biological Engineering Department, Colorado School of Mines, Golden, Colorado 80401, USA
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Ionic conductivity in a polymeric fuel cell requires water uptake. Previous theoretical studies of water uptake used idealized parameters. We report a parameter-free prediction of the water-swelling behavior of a model fuel cell membrane. The model polymers, poly(methyl-butylene)-block-poly(vinylbenzyl-trimethylamine), form lamellar domains that absorb water in humid air. We use the Scheutjens-Fleer methodology to predict the resulting change in lamellar structure and compare with x-ray scattering. The results suggest locally uniform water distributions. However, under conditions where a PVBTMA and water mixture phase-separate, the two phases arrange into stripes with a dilute stripe sandwiched between two concentrated stripes. A small amount of water enhances conductivity most when it is partitioned into such channels, improving fuel-cell performance. © 2015 AIP Publishing LLC. [http://dx.doi.org/10.1063/1.4914512]

I. INTRODUCTION

A. Alkaline-exchange membranes

Fuel cells oxidize fuel in a controlled way to efficiently convert the stored energy into electrical power. They work by performing the oxidation reaction in two steps: splitting fuel molecules at the anode and oxygen molecules at the cathode. In general, the purpose of the electrolyte is to promote transport of the reacting ions, inhibit transport of the opposite charges, and prevent cross-over of fuel or oxygen. Our goal is to develop electrolytes for consumer applications, similar to lithium-ion batteries, but with better energy density and rapid refueling.1 Low-temperature electrolyte types include aqueous alkaline, proton-exchange polymer membrane, and alkaline polymer membrane. The first two are well studied and have drawbacks; while alkaline exchange membranes solve these issues, more research is needed to improve their performance.

Aqueous alkaline electrolytes are highly efficient, but are impractical for most applications due to carbon dioxide susceptibility. For example, in KOH electrolytes, potassium carbonate crystals form, eventually clogging the system.2-4 In systems where it would be impractical to remove carbon dioxide, aqueous electrolytes are not feasible.

Proton-exchange membranes such as Nafion™ (DuPont)5 are more practical, but still suffer drawbacks. The two half-reactions do not occur readily in the acidic environment and therefore platinum is needed to catalyze the reactions; this is uneconomical. Also, platinum catalysts are poisoned by trace amounts of carbon monoxide, requiring carbon-free fuels. As an alternative, many in the community have recently turned their attention to alkaline fuel cells.6

Alkaline-anion-exchange-membrane (AEM) fuel cells can solve the problems of both liquid alkaline fuel cells and Nafion fuel cells. Due to their dimensional stability, they are not poisoned by CO2. Also, they can operate using a range of catalysts, such as nickel, since the reaction rates are much higher in an alkaline environment.3,7,8 These features allow the use of carbon-containing fuels such as methanol, instead of pure hydrogen and unprocessed air.1 Currently, however, Nafion fuel cells outperform their alkaline-membrane counterparts in power density: Nafion has achieved 720 mW/cm2 at 75 °C with Nafion-112,6 compared to, for example, 372 mW/cm2 at 50 °C with a PBI-KOH membrane.9 Therefore, more research is needed to make alkaline membranes suitable for consumer applications.

Commonly used ionic groups for AEMs include ammonium, phosphonium, and sulphonium.10,11 For this paper, we analyze a diblock copolymer where the hydrophilic block consists of polystyrene with a trimethyl ammonium side chain, poly(vinylbenzyl-trimethylamine)[Br−], referred to as PVBTMA. PVBTMA was first investigated as the hydrophilic block in an anion-exchange membrane by Zeng et al.12 The membrane had an ionic conductivity roughly 4 times lower than Nafion-115, but with roughly 10 times less methanol crossover. Therefore, the ratio of ion conductivity to fuel crossover was better than Nafion-115, which makes it a promising electrolyte material.

The research reported below is part of a team sponsored by the U.S. Army through a Multidisciplinary University Research Initiative (MURI). One of the main goals of the project is to study the fundamental properties of alkaline anion-exchange membranes and validate these predictions. Accordingly, this paper explores PVBTMA (Figure 2) joined to a hydrophobic polymer poly(methyl-butylene) to create a
lamella-forming diblock. The synthesis of the membranes is described elsewhere.13

B. Water distribution

1. Motivation

Hydroxide ion transport through the membrane is fundamentally dependent on the amount and distribution of water within the membrane, since the dominant ion transport mechanisms require water. Due to reasons not completely understood, ionic diffusion is roughly twice as fast for proton transport compared to hydroxide: $9.3 \times 10^{-9}$ m$^2$ s$^{-1}$ in 25°C water.14 Therefore, it is a challenge to make OH$^-$ fuel cells conduct as well as H$^+$ fuel cells.

For practical reasons, fuel cells are operated with relatively dry membranes. If ion conductivity was the only consideration, they would be much wetter.15 One problem with too much water is that water can flood the gas diffusion layers, the regions designed to allow the input gases to contact the catalyst.16 Another problem is that the membrane can become weak and develop holes, especially after repeated wet-dry cycles.17 Under these constraints, the way to maximize conductivity is to make the membrane heterogeneous on a microscopic scale, containing regions for structure and regions for ion diffusion. The ion-bearing regions absorb water, creating pathways for ion conduction. Examples include interpenetrating polymer networks18 and diblock copolymers,19 where each chain of hydrophilic polymer is connected to a hydrophobic polymer.

2. Morphology

The model poly(methyl-butylene)-blockpoly(vinylbenzyltrimethylamine) (PMB-b-PVBTMA) diblock studied here forms alternating layers of PMB and PVBTMA. Each connection point is constrained to lie at the interface between a PMB block and its adjoining PVBTMA block. PMB, a rubbery material also known as hydrogenated polyprene, was chosen as a compromise between structural rigidity and flexibility, allowing the membrane to maintain a lamellar structure while still being able to swell with water.

For simplicity, we focus on a lamellar morphology. The interface transition width and interface surface tension have been analyzed with field theory calculations, and depend on molecular weight and degree of repulsion.20,21 In our calculations, we ignore any effects of the PMB-PVBTMA interaction except to set the areal (grafting) density of the PVBTMA chains, as described in Sec. III C. We assume that PMB and PVBTMA are dissimilar enough that the interface between the two is flat up to the resolution of the calculation. The orientation of the morphology can be aligned using electric potentials or mechanical stress, minimizing the effect of tortuosity.22

3. Water channels

The results of this study suggest a new way to engineer morphologies beneficial to durability and ion transport. The type of morphology discussed in the previous paragraph does not depend on water absorption; membranes are usually cast dry, and the morphology depends on the hydrophilicto-hydrophobic polymer in this state. This dependence leads to using more hydrophilic polymer than would otherwise be ideal in order to establish a desirable morphology. We propose that an ideal membrane should have a low hydrophilic-to-hydrophobic ratio. Ion diffusivity in polyelectrolyte solutions increases exponentially or even faster with increasing water content than plateaus beyond roughly a 50% water mixture.13 Most fuel cell membranes do not reach such high water fractions, even within their conducting regions. This low water content is an intentional trade-off. Too much water causes the hydrophilic blocks to swell significantly and become gel-like, but this is only a problem if the hydrophilic polymers constitute a major portion of the membrane. In short, we would like to be able to use a small amount of hydrophilic polymer and yet still specify the morphology. Here is where water channels are important. Our model copolymer, for example, is composed of roughly equal parts PMB and PVBTMA, producing a well-connected lamellar morphology. If we could design the membrane to absorb water preferentially into narrow stripes down the center of each lamella, the rest of the PVBTMA block would remain dry and stable. Our preliminary results showed that this segregation can indeed occur if PVBTMA has a solubility gap, at the conditions that generate a two-phase coexistence.23,24 As an added benefit, we proposed that dissociated ions from the dense phase can leach into the dilute phase, enhancing conductivity even more. This turns out to be a negligible effect (see Appendix E), but could be exploited in future designs. After we proposed these ideas, there has been experimental evidence for lamellar water channels in a polystyrene-block-poly(styrene-sulfonate) membrane.25 One of the goals of this paper is to use numerical modeling to determine the properties of these water channels and under what conditions they form in a real system.

4. Theory background

Three classes of theory/simulation are used to study polymers: molecular dynamics, coarse-grained dynamics, and field theory.26 Of these, field theory is the best suited for finding the equilibrium spatial distribution of polymers in a dense solution, where thermal fluctuations are negligible. In this limit, all the methods agree, but field theory is the simplest and most direct.27 Polymer field theories are rooted in the similarity between a polymer’s partition function and that for a diffusing particle.28,29 Some ideal cases with symmetry can be studied analytically.29,30 Numerically, they can be studied using spectral methods31 or using real space methods. One of these, the Scheutjens-Fleer method,32,33 is based on weighted random walks on a lattice, with the lattice spacing set by the polymer’s flexibility according to the freely jointed chain model. The Scheutjens-Fleer method is equivalent to the discretized diffusion equation with numerical derivatives. Therefore, so-called continuum methods30,34 are equivalent, but may over-smooth features if the lattice spacing is too fine. Different lattice geometries produce equivalent results as long as the proper correction is made to the lattice spacing.35 These theories have been applied to two systems relevant to this work: polymer brushes and diblock copolymer lamellae.
5. Polymer brush background

Similar to an end-grafted polymer brush, the PVBTMA polymers in our system each have one end constrained to lie at the hydrophilic-hydrophobic interface, with the other end free. Polymer brushes are well studied. The first experiments measured the force between two facing polymer brushes immersed in a good solvent and poor solvent. Early theoretical work found the mean-field scaling behavior of the brush height or distance from the interface. In the limit of high molecular weight, the polymer profile was found to be parabolic. The scaling arguments were re-derived for charged polymer brushes in a strong salt solution and found to behave equivalently to neutral brushes in a good solvent. Numerical studies of two overlapping polymer brushes have mostly focused on ideal Flory-Huggins solvents with an interaction parameter, \( \chi \), near 0.5, i.e., the theta point, where the solvent is just poor enough to exactly cancel the polymer’s swelling due to self-avoidance, and the polymer’s self-interaction vanishes. The transition from poor to good solvent has been studied by molecular dynamics simulations and confirmed through neutron scattering.

6. This work in context

The aim of this paper is to use established polymer brush theory to study ionic copolymer membranes, with no free parameters. Experimental progress includes X-ray scattering on membranes with a selective solvent. Theoretical progress includes a mean-field calculation of polymer brush profiles, but most investigations assumed a Flory-Huggins equation of state near the theta point. This work is the first work that we are aware of that solves for the solvent distribution within a membrane with no free parameters. Specifically, we model the hydrophilic lamella as two facing polymer brushes and calculate how the lamella swells as a function of ambient humidity (water activity). We calculate the water profile within a lamella, since it is known from polymer brushes that water varies spatially due to competition between wetting and chain stretching. To calculate the spatial dependence, we solve the self-consistent field equations using the Scheutjens-Fleer method, a standard method for studying polymer brushes outlined in Sec. II C. The calculation requires three main inputs. First, the flexibility of the hydrophilic chains is inferred from literature values of similar polymers (Sec. III B). Second, the solubility of the hydrophilic block is measured for a sample of isolated hydrophilic block, not tethered to a hydrophobic block (Sec. III D). Third, the grafting density of hydrophilic block at the interface, which depends on humidity, is inferred from small-angle X-ray scattering (SAXS) (Sec. III C). Thus, all the parameters used for the calculation are experimentally determined. Finally, we attempt to verify the predictions as much as possible from independent experimental measurements.

7. Overview of this paper

Section IV shows the results of the calculations. Under most conditions (absent a phase separation), water is predicted to nearly uniformly absorb into the hydrophilic regions of PVBTMA. In that case, the overall membrane conductivity is straightforward to estimate. The amount of water absorbed determines the polymer concentration in the hydrophilic blocks. If one knows the conductivity as a function of concentration, then the overall conductivity can be extrapolated. If, however, there are distinct regions with more and less water, this simple calculation is no longer possible. As a concrete example, our calculations predict that under some conditions, water-rich slabs (“stripes”) may form at the center-plane of some hydrophilic lamellae. Heterogeneous water absorption is expected to conduct ions better than homogeneous absorption of the same amount of water, since conductivity increases faster-than-linearily with water content in this regime. Therefore, understanding the conditions that encourage uneven water absorption is important for developing fuel cells that conduct ions well without absorbing so much water that they lose integrity.

II. CALCULATION METHODOLOGY

A. Random walk representation

The Scheutjens-Fleer model is a discretized mean field theory defined on a lattice with Kuhn-length spacing (defined below). The mutual interaction of the monomers, including mutual avoidance, gives rise to a free energy cost for inserting a monomer at any position. This cost approaches a statistical average when the density of monomers is high. For closely packed chains, this is a good assumption, since each thermodynamically significant segment of polymer interacts with several others at any given time. All of our predictions are in the polymer-dense regime where the mean-field formulation applies. We study a lamellar geometry uniform in the \( x \) and \( y \) coordinates with variation only in \( z \) (perpendicular to the hydrophilic/hydrophobic interface, see Fig. 1).

At the scales considered, the specifics of the local chain configurations and their interactions become unimportant and each polymer can be represented as a random walk on a cubic lattice. The separation between lattice sites is chosen so that the random walks capture the real flexibility of the polymer at the theta point. That is, the lattice parameter is set to be the Kuhn length, \( \ell_K \), of the polymer (twice the persistence length). The Kuhn length is a measure of how coiled the polymer is at the theta point—the point at which the solvent is just poor enough to exactly cancel the polymer’s swelling due to self-avoidance. That is, a free polymer in dilute solution at the theta point behaves as a non-self-avoiding random walk with step lengths equal to the Kuhn length. To define it exactly, we require that:

1. the mean squared end-to-end distance matches the polymer at theta conditions: \( n \ell_K^2 = \langle r^2 \rangle_o \), where \( n \) is the number of segments, or Kuhn steps composing the random walk, and \( \langle r^2 \rangle_o \) is the average end-to-end span of the polymer at theta conditions.
2. the total length of the random walk is equal to the length of an ideal chain that is fully extended while still maintaining the covalent bond angle, \( \theta_b = 109.5^\circ \), of the hydrocarbon.
backbone: \( n t K = n_b \ell_b \sin \theta_0 / 2 \), where \( n_b \) is the number of backbone bonds, and \( \ell_b \) is the bond length (154 pm).

Combining the requirements gives

\[
\ell_K = \frac{\langle r^2 \rangle_0}{n_b \ell_b \sin \theta_0 / 2} = \frac{C_{\text{wo}} \ell_b}{\sin \theta_0 / 2},
\]

\[
n = \frac{n_b \sin^2 \theta_0 / 2}{C_{\text{wo}}},
\]

where \( C_{\text{wo}} \equiv \langle r^2 \rangle_0 / n_b \ell_b^2 \) (in the limit \( n_b \to \infty \)) is the experimentally measured characteristic ratio, as discussed in Sec. III B.

The calculation space, \( \vec{r} \), is discretized into a square Kuhn-length lattice. In particular, we introduce an integer variable \( \zeta \), which counts the number of Kuhn steps from the interface: \( z = \xi \ell_K \). Points within 1 Kuhn length of the interface are grouped together as \( \zeta = 1 \), points within 1 to 2 Kuhn lengths of the interface are \( \zeta = 2 \), etc. Let there be \( m \) lattice lengths between the interface and the lamella’s mid-plane so that \( z \) ranges from 1 to \( m \). This fixes the swelled hydrophilic lamellar width to \( 2m \ell_K \) (Appendix C) shows how to accommodate half-integer widths. The width depends on the amount of solvent admitted into the brush, as determined by the experimental conditions outlined below. Other functions such as \( \phi \) are similarly discretized. We may refer to \( \phi (z) \) and \( \phi (\zeta) \) interchangeably depending on context.

### B. Self-consistency

The following is a brief overview for the method of finding \( \phi_{\text{avg}} (\zeta) \), the mean-field polymer volume-fraction profile as a function of position. We start with an ansatz that the profile is given by some assumed \( \phi_{\text{avg}} (\zeta) \). A given Kuhn segment inserted at height \( \zeta \) interacts with \( \phi_{\text{avg}} (\zeta) \) via the chemical potential, \( \mu_K (\zeta) \). This potential depends only on the local average volume fraction, so \( \mu_K (\zeta) \) or \( \mu_K (\phi (\zeta)) \) may be used interchangeably. The functional form of \( \mu_K (\phi) \) can be obtained independently through experimental measurement, as discussed in Sec. III D, and we thus presume it to be known.

This is used here in a restricted sense to mean the energy required to insert a Kuhn segment into a uniform solution whose volume fraction is \( \phi_{\text{avg}} (\zeta) \); therefore, unlike the full chemical potential, \( \mu_K (\zeta) \) varies in space (from entropic effects due to the tethering of one end to the interface).

To test whether the ansatz \( \phi_{\text{avg}} (\zeta) \) is self-consistent, now consider a single test chain among the many. Every spatial configuration of the polymer (for example, see Fig. 1) exists with a certain probability, determined by \( \mu_K (\zeta) \) (to be spelled out in Sec. II C). By averaging all the possible configurations, we can calculate the test polymer’s average volume-fraction profile, \( \langle \phi (\zeta) \rangle \). If the solution is self-consistent, then the mean-field distribution equals the test-chain distribution

\[
\phi_{\text{avg}} (\zeta) = \langle \phi (\zeta) \rangle.
\]

In general, the first guess will not be self-consistent, but we can use an iterative process to converge to the self-consistent solution. Equation (3) is a set of \( m \) single-variable equations, one for each \( \zeta \). We solve this using the Newton-Raphson method,\(^{45}\) producing a self-consistent \( \phi_{\text{avg}} (\zeta) \). That is, we look for the root (or roots) of the function \( f (\zeta) \equiv \phi_{\text{avg}} (\zeta) - \langle \phi (\zeta) \rangle \) using the Jacobian \( J_{ij} = \partial f (\zeta) / \partial \phi_{\text{avg}} (\zeta) \). Note that \( \phi (\zeta) \) is a function of \( \phi_{\text{avg}} (\zeta) \) as spelled out in Sec. II C. Therefore, \( J_{ij} = \delta_{ij} - \delta (\phi (\zeta)) / \partial \phi_{\text{avg}} (\zeta) \), where \( \delta_{ij} \) is the Kronecker delta function. An iteration is \( \phi_{\text{avg}} (\zeta) \to \phi_{\text{avg}} (\zeta) - \lambda J_{ij} f (\zeta) \), where \( \lambda \in (0,1] \) as described in Numerical Recipes.\(^{45}\) We implemented the scheme in Mathematica. Each Newton step takes approximately 1 s to compute on a single, modest processor. It takes roughly 10–100 steps to converge to within 8 decimal places, with only occasional intervention required to nudge the algorithm away from a local minimum.

### C. Scheutjens-Fleer method

The following procedure calculates \( \langle \phi (\zeta) \rangle \) for the single test chain by efficiently summing up the Boltzmann weights for all the random walk configurations on the lattice. This is permissible since the lattice has Kuhn-length spacing; when considering the set of random walks the polymer can follow on such a lattice, its bending energy and self-interaction energy are implicitly included. Therefore, the only interaction that goes into the Boltzmann factors is \( \mu_K (\zeta) \), the interaction between polymer segment and the water-polymer solution at \( \zeta \). By construction, \( \mu_K (\zeta) \) includes the total free energy to insert a segment at \( \zeta \), including making space for it under osmotic pressure, or equivalently, displacing an equivalent volume of water. Therefore, the Boltzmann factor for each segment is just \( \exp [-\mu_K (\zeta) / k_B T] \). For a segment at position \( \vec{r} \), the coordinate is its non-dimensionalized projection onto the \( z \)-axis: \( \zeta = \vec{r} \cdot \hat{z} \ell_K^{-1} \). Given a 3-d Kuhn lattice random walk, \( \{ \vec{r}_j \} \), its relative probability of occurrence, \( R \), at fixed volume is simply the product of all the segment Boltzmann factors

\[
R (\{ \vec{r}_j \}) = \prod_{j=1}^{n} \exp [-\mu_K (\vec{r}_j \cdot \hat{z} \ell_K^{-1}) / k_B T].
\]
Since the lattice is Cartesian, there are of order $(2 \cdot 3)^n$ unique chain conformations. This can be reduced. Instead of finding the relative probability of a given $\hat{r}$, we can find the relative probability of a projection onto $\zeta$-coordinates, $\{\zeta\}$. $\{\zeta\}$ is similar to a 1-d random walk, but each coordinate can be $+1, 0,$ or $-1$, relative to the preceding coordinate. $A + \ell K \hat{z}$ step maps to $+1, -\ell K \hat{z}$ to $-1,$ and $+\ell K \hat{x}$ and $-\ell K \hat{y}$ all map to $0$. This leads to a redundancy factor, $\lambda$,

$$
\lambda(\zeta_j - \zeta_{j-1}) = \begin{cases} 1 & (\zeta_j - \zeta_{j-1}) = \pm 1 \\ 4 & (\zeta_j - \zeta_{j-1}) = 0 \end{cases}.
$$

(5)

The first Kuhn segment for all the chains occupies $\zeta = 1$ and so does not have an associated $\lambda$. The resulting expression for relative probability is

$$
R(\{\zeta\}) = \exp \left[ \frac{-\mu_K (1)}{k_B T} \right] \times \prod_{\zeta_{i-1}}^{\zeta_{n}} \lambda (\zeta_j - \zeta_{j-1}) \exp \left[ -\frac{\mu_K (\zeta_j)}{k_B T} \right].
$$

(6)

The inputs, $\{\zeta\}$, still have many combinations. The direct calculation for the resulting test-chain profile is given by summing over all chain conformations and binning their segment positions,

$$
\langle \phi (\zeta) \rangle \propto \sum_{\{\zeta\}} R(\{\zeta\}) \sum_{i} \delta_{\zeta_i, \zeta},
$$

(7)

where $\delta$ is the Kronecker delta function. Scheutjens and Fleer discovered a computational trick to calculate $\langle \phi \rangle$ in only order $m \cdot n$ computations. It makes use of two Green’s functions. First, $G_{i \rightarrow i}^{\zeta}$, defined as the sum of partial Boltzmann products for segment $i$ through segment $i$ (i.e., $R(\{\zeta\})$ in Eq. (6) with $n = i$) for all possible chain conformations beginning at the interface ($\zeta = 1$) and ending with segment $i$ in layer $\zeta$. $G_{i \rightarrow n}^{\zeta}$, defined as the sum of partial Boltzmann products for segment $i$ through final segment $n$ for all conformations with segment $i$ in Kuhn cell $\zeta$ (the “*” indicates that the final position is unspecified). Then, by construction, the probability of finding segment $i$ in cell $\zeta$ is

$$
P_i (\zeta) \equiv \alpha G_{i \rightarrow i}^{\zeta \rightarrow \zeta} / \exp [ -\mu_K (\zeta) / k_B T],
$$

(8)

where $\alpha$ is a normalization factor. The divisor is to account for the fact that the $i$th Boltzmann factor is counted in both Green’s functions. The normalization factor, $\alpha$, can be determined by requiring $P_i (1) = 1$; that is, all the 1st segments originate at the interface and exist in the 1st Kuhn cell. The efficiency trick comes from the fact that the Green’s functions can be generated inductively. The process is illustrated in Appendix B. Starting with the distribution of segment 1, assumed to all be attached at the interface,

$$
G_{1 \rightarrow 1} = \begin{cases} \exp [ -\mu_K (1) / k_B T ] & \zeta = 1 \\ 0 & \zeta > 1 \end{cases}.
$$

We may build expressions for $G$ involving further steps along the chain inductively,

$$
G_{i \rightarrow i+1}^{\zeta} = \exp [ -\mu_K (\zeta) / k_B T ] \times \begin{cases} 0 & \zeta < 1 \\ G_{i \rightarrow i}^{\zeta-1} + 4G_{i \rightarrow i}^{\zeta} + G_{i \rightarrow i}^{\zeta+1} & 1 \leq \zeta < m \\ G_{i \rightarrow i}^{\zeta-1} + 5G_{i \rightarrow i}^{\zeta} & \zeta = m \end{cases}.
$$

(9)

The overall prefactor is the Boltzmann factor associated with the $(i + 1)$th Kuhn segment. The first case gives the effect of the (assumed) impenetrable diblock interface at $z = 0$. The second case applies to the bulk of the lamella, away from the boundaries. The terms arise from the different ways in which the random walk can arrive at $\zeta$ after $i$ steps. Relative to the $(i + 1)$th segment, the $i$th segment can either be $+z$ (first term), $\pm x$ or $\pm y$ (second term), or $-z$ (third term). The third case creates a mirror boundary condition at $z = m \ell K$; if the $(i + 1)$th segment would cross the mirror, it is instead reflected back. The mirror is on the face of the $\zeta = m$ cell, so if the $(i + 1)$th segment is in the $+z$ direction, it starts and ends at $\zeta = m$. This, plus the $\pm x$ and $\pm y$ steps, gives the factor of 5. A variation of this boundary condition is shown in Appendix C. This representation is equivalent to evaluating two sets of polymer brushes, one on the left interface and one on the right, overlapping in the middle. To get the polymer distribution in the full lamella, the left profile is simply reflected about $z = m \ell K$.

The second set of Green’s functions needs to be calculated starting from the free end of the chain and working back toward the tethered end: $G_{n \rightarrow n}^{\zeta} = \exp [ \mu_i (\zeta) / k_B T ]$, and $i$ is iterated downward,

$$
G_{(i-1) \rightarrow n}^{\zeta} = \exp [ -\mu_K (\zeta) / k_B T ] \times \begin{cases} 0 & \zeta < 1 \\ G_{i \rightarrow n}^{\zeta-1} + 4G_{i \rightarrow n}^{\zeta} + G_{i \rightarrow n}^{\zeta+1} & 1 \leq \zeta < m \\ G_{i \rightarrow n}^{\zeta-1} + 5G_{i \rightarrow n}^{\zeta} & \zeta = m \end{cases}.
$$

(10)

For each position $\zeta$, the polymer density due to the Boltzmann-weighted configurations can be calculated by adding up the contributions due to every $i$th segment from $i = 1$ to $n$, and normalizing

$$
\langle \phi (\zeta) \rangle = \left( \frac{\sigma V_K}{\ell K} \right) \sum_{i=1}^{n} P_i (\zeta).
$$

(11)

The prefactor in parentheses is the fraction of volume that the initial segments collectively occupy within the first lattice cell, and depends on the average volume of a Kuhn segment, $V_K$; the Kuhn length, $\ell K$; and the grafting density, $\sigma$, defined as the number of PVBTMA attachments divided by the interfacial surface area. This equation is used to compute the iterative refinements outlined in Sec. II B that result in a self-consistent $\phi_{\text{avg}}(\zeta)$.

In summary, this method calculates the polymer distribution for a specific lamellar width $2m \ell K$. In experiments, however, the control variable for swelling is the water activity, $a$ (the ambient relative humidity with which the membrane is in equilibrium). To relate the two, we note that along the center plane of the calculation, $z = m \ell K$, the potential is symmetric and so the polymers behave locally as random
walks. Hence, they cannot support an osmotic pressure, and so the chemical potential of water at the center plane is the same as that of a water molecule in an un tethered polymer solution of concentration $\phi$. This is the same $\mu_w(\phi)$ determined in Sec. III D 1. That is, the water activity of the calculation output, $\phi_{\text{avg}}(\zeta)$, is given by

$$a = \exp \frac{\mu_w(\phi_{\text{avg}}(m))}{k_B T}.$$  

(12)

### D. Calculation of $\mu_K$

Section II C requires the work, $\mu_K(\zeta)$, to take an isolated Kuhn segment and insert it into layer $\zeta$ at fixed volume. Evidently, this only depends on the local polymer isolated Kuhn segment and insert it into layer $\zeta$ at fixed volume. Evidently, this only depends on the local polymer volume fraction, $\phi$, so we seek the function $\mu_K(\phi)$, and $\mu_K(\zeta) = \mu_K(\phi(\zeta))$. The quantity $\mu_K$ must be inferred from experiment (Sec. III D), which measures the chemical potential of water, $\mu_w(\phi, p_0)$, as a function of concentration, $\phi$, at ambient pressure $p_0$. Consider a large volume, $V$, of solution at volume fraction $\phi$, pressure $p$, and temperature $T$. For convenience, we divide the volume into $N$ units with volume equal to the average volume, $V_{H_2O}$, of a water molecule. Of these, $N_s = (1 - \phi) N$ consist of water and $N_i = \phi N$ consist of polymer (we use the subscript $s$ since the argument is applicable to any solute, not just polymers). In terms of the Helmholtz free energy, $A, \mu_w$ is defined by

$$\mu_w = \left( \frac{\partial A}{\partial N_w} \right)_{N_s, V, T}.$$  

(13)

The derivative is performed holding fixed $N_s$, volume $V$, and temperature $T$. We need to find $\mu_s$, the free energy cost to add a unit of polymer. Since the process of adding a Kuhn segment is isotropic, the Gibbs-Duhem equation simplifies to $N_s d\mu_s = -N_w d\mu_w + V dp$. In the limit where the layer is much larger than a single unit, $dp \to 0$. Therefore,

$$N_s d\mu_s = -N_w d\mu_w.$$  

This equation can be integrated to give $\mu_s(\phi, p_0)$, using $(1 - \phi) N_s = \phi N_s$. Holding pressure constant at $p_0$ and using a dummy integration variable, $\bar{\mu}_w$,

$$\mu_s(\phi, p_0) = - \int_{-\infty}^{\phi} \frac{\mu_w(N_w)}{N_s} d\bar{\mu}_w.$$  

(14)

Transforming the integration variable to $\phi$, and using the dummy variable $\bar{\phi}$,

$$\mu_s(\phi, p_0) = - \int_{1}^{\phi} \frac{1 - \bar{\phi} \mu_w(\bar{\phi}, p_0)}{\phi} d\bar{\phi}.$$  

Integrating by parts,

$$\mu_s(\phi, p_0) = \phi - 1 \phi \mu_w(\phi, p_0) - \int_{1}^{\phi} \frac{\mu_w(\phi, p_0)}{\phi^2} d\phi.$$  

(15)

The pressure dependence can be found from one of the Maxwell relations,

$$\left( \frac{\partial \mu_s}{\partial p} \right)_{T, N_j, N_j} = \left( \frac{\partial V}{\partial N_i} \right)_{T, p, N_j, N_j}.$$  

The right side of this equation is just $V_{H_2O}$, since the unit volume for each species is taken to be the same. Therefore, both pressure derivatives are equal,

$$\left( \frac{\partial \mu_s}{\partial p} \right)_{T, N_i, N_j} = \left( \frac{\partial \mu_w}{\partial p} \right)_{T, N_i, N_j}.$$  

Integrating with respect to $p$ gives

$$\mu_s(\phi, p) - \mu_s(\phi, p_0) = \mu_w(\phi, p) - \mu_w(\phi, p_0).$$  

Using Eq. (14) and noting that at equilibrium, $\mu_w(\phi, p)$ is constant between layers, gives the solute chemical potential up to a constant

$$\mu_s(\phi, p) = -\frac{\mu_w(\phi, p_0)}{\phi} - \int_{1}^{\phi} \frac{\mu_w(\phi, p_0)}{\phi^2} d\phi$$

$$= \mu_s(\phi).$$  

(16)

We have omitted the constant since it does not affect the Scheutjens-Fleer calculation. Finally, $\mu_K$ is simply the free energy required to insert $V_{H_2O}$, of a water molecule. Of course, $\mu_K$ includes the free energy of the water per chain, since $\mu_K$ is thermodynamically favored. It turns out that $\mu_K$ can also be used to compare between two profiles with different amounts of water, since
the hypothetical water reservoir (not the humid air) is at zero chemical potential: \( \mu_w(\phi = 0) = 0 \). This can be seen more clearly by introducing a hypothetical intermediary state, one in which two artificial partitions are introduced that segregate a layer of pure water in the middle of the lamella. The energy to introduce the partition is implicitly counted in \( A \) for the partitioned system, since the available system states are constrained. Then, if water is added to the center swelling the lamella, this change happens with no change in free energy, since the added water does not affect the chains outside of the partitions. When the partitions are released, the chains assume their self-consistent form, and a new \( A \) can be calculated and directly compared with the previous \( A \).

Our calculation also assumes a variable chain areal density on the interface, \( \sigma \), which the system sets by minimizing the free energy associated with the interface surface (Sec. III C). Unfortunately, \( \sigma \) is inferred directly from the experiment so the underlying free energy contributions are unknown. Despite this missing term, which is presumed to be small for our system, we proceed to use \( A \) to distinguish between multiple solutions at the same width and/or water activity.

Care should be taken regarding discretization of the spatial coordinates. In some cases, the Scheutjens-Fleer method produces a fake solution that is artificially self-consistent in a discretized framework, but would not be self-consistent in a continuous framework. This effect occurs, for example, if there is a sudden jump in \( \phi \), such as at the interface of a water channel (Sec. IV). At such an interface, there is a strong energy barrier between the two phases; a point in the dilute phase cannot easily move to the concentrated phase, and vice versa. In effect, the discretization adds an implicit, unphysical force to the system that corrupts \( A \). To find the correct water channel width, we used an annealing process. We modulated the interaction strength via \( V_k/V_{H_2O} \) until the profile was smooth, and then gradually escalated back to the correct \( V_k/V_{H_2O} \). This process ensured a correct water channel width.

### III. Measurement of Parameters

In order to calculate the polymer distribution within a hydrophilic layer of the PMB-PVBTMA membrane sample, several parameters describing these polymers are needed. Sec. III A highlights pertinent synthesis and characterization details for the PMB-b-PVBTMA membrane sample. Section III B estimates the Kuhn length for PVBTMA, a measure of its flexibility. Sec. III C describes x-ray and water uptake experiments used to determine the polymers’ “grafting density” at the block interface. Finally, Sec. III D details water uptake experiments on isolated, untethered PVBTMA, resulting in an equation of state.

#### A. Synthesis Characterization

For each block of the membrane, we need the degree of polymerization and the average volume of a monomer.

Gel permeation chromatography and nuclear magnetic resonance spectrometry experiments determined that each hydrophilic block comprises on average 160 repeat units, 84% of which are PVBTMA[Br\(^-\)], and the remainder are poly-(4-methylstyrene). Each hydrophilic block comprises on average 358 repeat units of PMB, with an isomer ratio, (1,4):(1,2):(3,4), measured to be 0.895:0.042:0.064 by mass. The polymers are relatively uniform with a polydispersity of 1.11.\(^1\)

The density of PVBTMA[Cl\(^-\)] is 1.325 g/cm\(^3\),\(^47\) from which we estimate the density of PVBTMA[Br\(^-\)] to be 1.60 g/cm\(^3\),\(^4\)\(^5\) taking account of the larger mass and larger displaced volume of Br\(^-\) compared to Cl\(^-\). The un-brominated monomers are poly-(4-methylstyrene), which has a density of 1.106 g/cm\(^3\). Taking into account the 84:16 mole ratio and assuming ideal mixing,\(^48\) this results in an estimated average monomer volume of 0.254 nm\(^3\), and block molecular weight of 37 400 amu.

For PMB, a similar isomer blend found in the literature has a density of 0.853 g/cm\(^3\).\(^49\) This results in an average monomer volume of 0.136 nm\(^3\) and block molecular weight of 25 060 amu. For reference, this blend of PMB has a glass transition temperature of \(-63°C\),\(^50\) so it is indeed rubbery and able to allow the membrane to swell.

#### B. Kuhn Length

The Kuhn length, \( \ell_K \), has not been determined for our PVBTMA polymer, but the value can be bounded using reported measurements of the characteristic ratio, \( C_{\infty} \), for related polymers. PVBTMA[Br\(^-\)] (Figure 2(d)) can be thought of as polystyrene (Figure 2(a)) with an added trimethyl group and an electrolyte. Polystyrene has a characteristic ratio of 9.85.\(^50\) The trimethyl group adds a steric effect, increasing the characteristic ratio. For example, poly-(4-tert-butylstyrene) (Figure 2(c)) is sterically similar and has a characteristic ratio in the range of roughly 11.9–13.4.\(^50\) Adding electrolytes to a polymer also tends to increase the characteristic ratio in a salt-free solution, but this effect is vanishingly small for high polymer concentration.\(^51\) The increase in characteristic ratio, \( \Delta C_{\infty} \), due to adding an electrolyte to each monomer can be estimated using the following Odtik expansion\(^52\) for the relative increase in Kuhn length, \( \Delta \ell_K \). For example, consider polystyrene at 50% concentration in water at 60°C. With \( C_{\infty}^{(PS)} = 9.85 \), this implies a Kuhn length of \( \ell_K^{(PS)} = 1.86 \) nm (Eq. (1)), a segment comprising 7.38 monomers and occupying volume \( V_k^{(PS)} = 1.27 \) nm\(^3\). The number charge density in this scenario is \( n_q = 0.50 \times (1.27 \text{ nm}^3)^{-1} \). Then, the relative increase in the characteristic ratio, \( \Delta C_{\infty}/C_{\infty} \), is

\[
\frac{\Delta \ell_K^{(PS)}}{\ell_K^{(PS)}} = \frac{\lambda_D^2/2Q}{\ell_K^{(PS)}} = \frac{\pi(ek_BT)^2}{n_q e^2 \ell_K^{(PS)}} = 0.0132,
\]

a negligible increase. Here, \( \lambda_D = \sqrt{ek_BT/2 \pi n_q e^2} \) is the Debye length, \( Q = e^2/\kappa_T k_BT \) is the Bjerrum length, and \( \kappa_T \) is the elementary charge. We used a permittivity, \( \epsilon = 35 \epsilon_0 \), roughly between that of water and dry polystyrene.

Taking the steric and dielectric effects into account, \( C_{\infty} \) for PVBTMA should be within the range of 9.85–13.4 or slightly higher. We experimented with values in this range and found no significant effect on the resulting \( \phi(\xi) \) profiles. This insensitivity suggests that polymer stretching effects play a minor role in our system. This is further corroborated by the
results, Sec. IV, where we use $C_\infty = 9.85$ as in polystyrene, with the understanding that the calculation may overestimate stretching effects. Similar to polystyrene, a Kuhn segment of PVBTMA is 1.86 nm long and 7.38 monomers. Each chain has 160 repeat units of PVBTMA, and therefore 22 Kuhn steps.

An important quantity for the Scheutjens-Flory calculation is the number of water molecules displaced by a Kuhn segment. The chain is composed of 83% [PVBTMA][Br$^-$], a single repeat unit of which displaces 0.265 nm$^3$. The remainder is poly(4-methylstyrene), which displaces 0.193 nm$^3$. Therefore, $V_K = 1.87$ nm$^3$ and $V_K/V_{H_2O} = 63$, that is, each Kuhn segment displaces 63 water molecules. Because $V_K/V_{H_2O} \gg 1$, any non-uniformity in the polymer chemical potential is strongly penalized; for each polymer segment available to take a random step, there are 63 water molecules available to level gradients in chemical potential.

**C. Chain areal density, $\sigma$**

X-ray scattering of the PMB-PVBTMA membrane is used to calculate the grafting density as a function of water activity, $\sigma(a)$, needed for Eq. (11). The grafting density depends on water activity: unlike in a traditional surface-grafted polymer brush, the chains have the freedom to spread laterally, at the expense of added surface energy.$^{54,55}$ To compute $\sigma$, one must know the layer width for both PMB and PVBTMA. The combined width is straightforward to measure using SAXS, where it is known as the lamellar repeat spacing, $d$. It is determined from the wave number at the primary peak, $q^*$, as $d = 2\pi/q^*$. These measurements were performed on the membrane sample at the Advanced Photon Source at Argonne National Laboratory. Azimuthally averaged scattering intensity profiles are shown in Figure 3(a) for the five humidity levels sampled.$^{13}$ For each humidity run, peaks are visible at $1q^*, 2q^*$, and $3q^*$, indicative of lamellar geometry. Note that some peaks are diminished in size due to destructive interference: if $k$ is the hydrophilic layer width divided by $d$, then we should expect to not see the $k$th order peak when $kd$ is an integer.$^{56}$ In this case, the 2nd order peak nearly vanishes for $a = 0.901$ and the 3rd order peak nearly vanishes for the lower humidity curves. Fitting the data with a 3rd-order Hermite interpolation determined $q^*$ to within about 0.04%. A change in $q^*$ indicates change in $d$-spacing, shown in Figure 3(b), given that the peaks do not change shape. It begins at 38.9 nm for the dry membrane, then drops down slightly and remains lower for the next three humidity steps. All the while, the membrane is swelling with water, indicating that the swelling is predominant by decreasing $\sigma$. In the last humidity step, $d$ increases to 42.8 nm, predominantly through widening of the PVBTMA block.

To determine the proportion of the lamellar spacing occupied by each block, one possible method is to fit the scattering profiles to simulated scattering curves. With only a few discernible peaks in the data, however, we were unable to get accurate results.

Instead, mass uptake can be used to determine the block-width ratio. A DVS run was performed using the same membrane as in the SAXS experiment, under identical conditions and humidity stepping protocol. The DVS determined how much water was absorbed after each humidity step. Assuming ideal mixing of PVBTMA and water, this determines the swelling of the PVBTMA layer. Consider a single polymer in the membrane. Let $v_1$ (48.7 nm$^3$) be the volume of its PMB block, $v_2$ (40.6 nm$^3$) be the volume of its PVBTMA block, and $v_w$ be the average volume of water per chain in the membrane, as measured from DVS. Then, the total volume per chain is $v_{\text{tot}} = v_1 + v_2 + v_w$. This quantity can also be expressed as

![FIG. 2. Chemical structure of polymers similar to PVBTMA with known characteristic ratios, $C_\infty$. (a) Polystyrene, 9.85. (b) Sodium poly(styrene-sulfonate): the charged group increases $C_\infty$ slightly due to screened charge repulsion. (c) Poly(4-tert-butylstyrene), 11.9–13.4: the trimethyl group increases $C_\infty$ due to steric repulsion. (d) PVBTMA: its $C_\infty$ should be within the range 9.85–13.4. We used $C_\infty = 9.85$ in the calculation.](image)

![FIG. 3. (a) Log–log plot of scattering intensity vs. wave number, $q$, showing peaks characteristic of lamellar structure. The smallest $q^*$ peak for each water activity is used to calculate the lamellar repeat length, $d = 2\pi/q^*$. (b) This $d$-spacing (middle curve), combined with dynamic vapor sorption (DVS) mass uptake of the membrane sample (top curve), gives the grafting density, $\sigma$. (bottom curve) versus relative humidity using Eq. (18). The linear interpolation of $\sigma$ is needed in Eq. (14).](image)
TABLE I. Quantities inferred from scattering and DVS experiments as inputs to brush calculation. The dimensionless quantity \( \sigma V_K / \ell_K \); grafting density times Kuhn segment volume by length, is the prefactor required in Eq. (11). The other rows from top to bottom are lamellar spacing, \( d \), measured from SAXS; volume of absorbed water per chain, \( v_w \), determined from DVS; and grafting density, \( \sigma \), determined from Eq. (18).

<table>
<thead>
<tr>
<th>( a )</th>
<th>0.032</th>
<th>0.271</th>
<th>0.515</th>
<th>0.742</th>
<th>0.901</th>
</tr>
</thead>
<tbody>
<tr>
<td>( d ) (nm)</td>
<td>38.9</td>
<td>38.4</td>
<td>38.3</td>
<td>38.3</td>
<td>42.8</td>
</tr>
<tr>
<td>( v_w ) (nm(^3))</td>
<td>0.354</td>
<td>3.03</td>
<td>6.23</td>
<td>10.2</td>
<td>17.1</td>
</tr>
<tr>
<td>( \sigma ) (nm(^{-2}))</td>
<td>0.217</td>
<td>0.208</td>
<td>0.200</td>
<td>0.192</td>
<td>0.201</td>
</tr>
<tr>
<td>( \sigma V_K / \ell_K )</td>
<td>0.219</td>
<td>0.210</td>
<td>0.202</td>
<td>0.194</td>
<td>0.203</td>
</tr>
</tbody>
</table>

\( v_{\text{tot}} = \sigma^{-1} d/2 \), since \( \sigma^{-1} \) is the average interfacial area per chain. Combining these two expressions,

\[
\sigma = (d/2)(v_1 + v_2 + v_w)^{-1}. \tag{18}
\]

The normalization constant, \( \sigma V_K / \ell_K \), is derived from the values in Table I for the five water activities studied in the SAXS experiment. The values remain within a narrow range of 0.194 to 0.219. Nevertheless, we include this variation in the brush calculation by defining the function \( \sigma(a) \) as the linear interpolation of the measured values (see Figure 3).

That is, the water activity determined by the brush profile via Eq. (12) must equal the water activity used to determine \( \sigma(a) \) in Eq. (11). This additional requirement for self-consistency can be incorporated by using \( \phi_{\text{avg}}(m) \) to calculate \( a \) (Eq. (12)), and then using that \( a \) to interpolate \( \sigma \) for use in Eq. (11). Appendix D shows that except for the uptick of the last point, the measured \( \sigma(a) \) is consistent with scaling theory.

**D. PVBTMA water absorption vs. \( a \)**

To account for the interaction between water and PVBTMA in solution, the model requires the chemical potential \( \mu_K(\phi) \) of the loose hydrophilic polymer (i.e., just PVBTMA with free ends). This requirement was simplified in Sec. II D to finding the chemical potential of water, \( \mu_w \), in such a sample. We explain the process of determining \( \mu_K(\phi) \) in the following: Sec. III D 1 describes the dynamic vapor sorption experiment, which measures the total swelled mass, \( M \), of the sample as a function of water activity, \( a \). Sec. III D 1 introduces a phenomenological function used to interpolate and extrapolate \( \mu_w(\phi) \) based on generalizing the Flory-Huggins expression to arbitrary polynomial order. Sec. III D 2 details how \( \mu_w(\phi) \) is transformed to the experimental variables and fit to the data, and the model selection process. Section III D 3 highlights three fits that are supported by the data, two of which are nearly identical and indicate that the sample is insoluble below \( \phi \approx 0.5 \). The other fit indicates a phase separation between two phases, \( \phi = 0.23 \) and \( \phi = 0.81 \). Both of these possibilities are explored in the Schuetjens-Fleer calculation in Sec. IV.

1. **Experimental measurement of \( a(M) \)**

A DVS apparatus, DVS Advantage (Surface Measurement Systems, Ltd.), was used\(^13\) to measure the mass, \( M \), of the swelled PVBTMA sample at controlled temperature, \( T = 60 \degree C \), and water activity, \( a \). This gives the activity of the sample as a function of its total mass, or \( a(M) \) (see Figure 4).

Water-soluble hydrophilic block samples were dissolved in water at approximately 40% by weight and pipetted onto the surface of the sample tray. The diblock membrane samples were cut to approximately 2 mm × 2 mm and placed on the tray. The DVS experiment was conducted at a constant temperature (60 \degree C) and various humidity set points. The following sequence was repeated three times and the results averaged. Each run starts with 6 h of dry nitrogen flow to remove the initial water, followed by a humidity stepping sequence: 10%, 20%, 30%, 40%, 50%, 55%, each for 120 min; 60%, 65%, 70%, 75%, 80%, 85%, 90%, each for 80 min; and 95% for 140 min. The sequence was chosen to ensure satisfactory equilibration: the decrease of the mass uptake following the step was well described by an exponential decay. We equilibrated at each step for at least five times the exponential decay time. This procedure does not rule out the possibility of additional, much slower equilibration processes. Thus, all our measurements should be viewed as applicable to materials equilibrated on an hour time scale.

Due to differences in synthesizing the PVBTMA in the membrane vs. stand-alone, the polymers are slightly different. The polymer used in this section is in chloride form (random 84% by mole PVBTMA[Cl]-, 16% P4MS), with degree of polymerization 277. To be consistent despite these differences, we deal with \( \mu_w(\phi) \), the chemical potential per water molecule in solution with polymer at volume fraction \( \phi \). Non-dimensionalizing in this way allows the halides to be interchangeable (see Appendix A).

a. **Generalized Flory-Huggins \( \mu_w(\phi) \)** We need a parameterized phenomenological function to interpolate the DVS data that capture the required behavior in the \( \phi \rightarrow 1 \) limit as well as the \( \phi \rightarrow 0 \) limit, which our DVS data do not access. The following generalized Flory-Huggins equation\(^57\) has the...
proper limiting behavior and fits the data well:

$$\mu_w(\phi)/(k_B T) = \log(1 - \phi) + \sum_{i=1}^{D} c_i \phi^i. \tag{19}$$

In the canonical formulation, using uncharged polymers, $D = 2$, $c_2$ is called $\chi$, and $c_1 = 1 - r^{-1}$, where $r$ is the volume ratio of a PVDBMA block to a solvent molecule (here, $r = 63 \cdot 24 > 1$). The deviation of $c_1$ from unity is due to the polymer’s center-of-mass translational entropy. We ignore this effect, since it is tiny in the loose polymer sample and absent in the membrane. Polyelectrolytes, however, have another source of entropy from dissociated counter-ions, an effect which can be estimated. The Bjerrum length, $Q$, for the polymer is calculated to be $Q = 0.752 \text{ nm}$, while the average charge spacing along the backbone, $A$, is calculated to be $A = 0.211 \text{ nm}$. The ratio $Q/A = 3.55 > 1$, which means charge screening dominates, thus inducing counter-ion condensation and increasing the distance between uncondensed ions from $A$ to approximately $Q$. Therefore, in dilute mixtures, $\mu_w(\phi)/(k_B T) \approx -(A/2Q) (\phi/\rho_w) V_{H,O} \approx -0.0166 \phi$, where $V_{p_w} \approx 0.254 \text{ nm}^3$ and $V_{H,O} = 0.0299 \text{ nm}^3$ are the average volume of a monomer and of a water molecule, respectively. This fixes $c_1 = 0.9834$. Although the calculation leading to this value is only a rough estimate, it shows that $c_1$ cannot deviate far from unity. The higher-order coefficients are determined by fitting.

2. Fitting $\mu_w(\phi)$ to $a(M)$

Equation (19) gives $\mu_w(\phi)$, but the DVS data are of the form $a(M)$. Therefore, we need to find $\phi(M)$ and $\mu_w(a)$, and perform least-squares fitting using the function $a(\mu_w(\phi(M)))$. Finding $\phi(M)$ requires the density of water, $\rho_w = 1.00 \text{ g/cm}^3$, and that of the hydrophilic polymer in chlorform at 83.9% functionalization, estimated as $\rho_p = 1.29 \text{ g/cm}^3$. The dry polymer mass, $M_p$, becomes a fitting parameter. Assuming ideal mixture,

$$\phi(M) = (M_p/\rho_p)/(M_p/\rho_p + (M - M_p)/\rho_w), \tag{20}$$

since the absorbed water mass is $(M - M_p)$. The second relation, $a(\mu_w)$, is found from assuming ideal gas, which is a good approximation at ambient pressure: $\mu_w = k_B T \log a$.

Figure 4 shows the least-squares fits for $D = 2$ through $D = 6$, and Table II lists their parameters, $\{M_p, c_2, \ldots, c_D\}$.

Except for the $D = 2$ fit, the curves qualitatively follow the data and improve with increasing $D$, until $D = 6$ which practically overlaps with $D = 5$. This is not surprising since the set of functions of order $D$ is a superset of the functions of order $D - 1$. Without care, however, this can lead to over-fitting the data. We use the corrected Akaike Information Criterion, or AICc, to compare best fits of each order. The AICc is given by $J \log RSS/j + 2K + 2(K + 1)/(J - K - 1)$, where $J = 45$ is the number of data points, and $K = D + 1$ is the number of model parameters (i.e., $\{M_p, c_2, \ldots, c_D\}$ plus $\sigma^2$, the estimated variance). Only differences in AICc matter. The model with the lowest AICc, in this case $D = 5$, is the likeliest. The difference, $\Delta AICc$, between a given model’s AICc and the smallest AICc, determines the model’s relative likelihood, which is given by $\exp(-\Delta AICc/2)$. Table II shows the RSS, $\Delta AICc$, and normalized likelihood, $L$, for each of the models considered. The highest likelihood is for $D = 5$ at 81.1%, followed by $D = 6$ at 18.7%, and $D = 4$ at 0.12%. $D = 2$ and $D = 3$ are ruled out, with negligible likelihood ($<10^{-13}$). We stopped at $D = 6$ since AICc will continue to rise thereafter, indicating over-fitting.

3. Two distinct models, $D = 4$ and $D = 5$

It turns out that the fit for $D = 4$ differs significantly from $D = 5$ and $D = 6$, despite all three being close fits to the DVS data. This is largely due to the difference in fitted $M_p$, to which the model is very sensitive, since it shifts the position of the logarithmic divergence in $\mu_w$. Also, there is no data below about $\phi = 0.74$, so the models must extrapolate there. The discrepancy between the different fits can be seen in plots of their Helmholtz free energy per lattice cell, $A/N$, vs. $\phi$ (see Figure 5), found by integrating Eq. (13). In regions where the curve is purely convex, such as at high $\phi$, the two components form a stable, uniform mixture. Where the curve is concave, the system prefers to follow the common tangent line (dashed, e.g.), the line connecting two points $(\phi_X, A(\phi_X))$ and $(\phi_Y, A(\phi_Y))$ where $A'(\phi_X) = A'(\phi_Y)$. These two conditions are equivalent to $\mu_w(\phi_X) = \mu_w(\phi_Y)$ and $\mu_p(\phi_X) = \mu_p(\phi_Y)$, meaning that solutions at $\phi_X$ and $\phi_Y$ coexist stably. A point along the common tangent line signifies that the sample is phase-separated, with part at $\phi_X$ and part at $\phi_Y$. For future reference, one should consider using this modified form of the curve to perform the least-squares fitting, especially if the sample is observed to phase separate. Otherwise, a fit could not be accepted.

<table>
<thead>
<tr>
<th>$D$</th>
<th>$M_p$</th>
<th>$c_2$</th>
<th>$c_3$</th>
<th>$c_4$</th>
<th>$c_5$</th>
<th>$c_6$</th>
<th>RSS</th>
<th>$\Delta AICc$</th>
<th>$L$</th>
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<tr>
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<td>0.00990</td>
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</table>
imply a solubility gap where none is seen in the data. We used the unmodified curve, which is suitable if the sample enters the metastable region, a possibility we could not rule out. This ambiguity is only an issue for the $D = 4$ fit.

Figure 5 shows that even though the fit for $D = 4$ approximates the DVS data reasonably well, it suffers from the inconsistency mentioned in the previous paragraph. According to the fit, the sample should phase separate between $\phi = 0.23$ and $\phi = 0.81$. The data, though, seem to exclude the top of this range since measurements were taken as low as $\phi = 0.74$. It is possible, though, that the mixture was either in a metastable state or barely phase separated—its appearance changed from matte to glossy. Therefore, we cannot rule out the $D = 4$. Unfortunately, standard DVS protocol is to halt the experiment at the first signs of phase separation to avoid flooding the tray. For future experiments, we would recommend carefully determining the composition of the coexisting phases. Even if the $D = 4$ suffers from some inconsistencies, it makes for a good contrasting example to the other two.

Figure 5 shows that the fits $D = 5$ and $D = 6$ are thermodynamically practically identical to one another. They form a single-phase solution between $\phi = 0.72$ and $\phi = 1$, dry polymer. There is a solubility gap below $\phi = 0.72$, as indicated by the dashed line. The gap extends down to a stable phase of ultra-pure water, $\phi \approx 0.06$, where the curve dips below zero before turning back to intersect the origin (not visible). The DVS did not probe below about $\phi = 0.74$, since the sample showed signs of phase separation. This is consistent with the phase separation predicted by the $D = 5$ and $D = 6$ models. In addition, a slightly altered sample with fewer ionic groups showed strong immiscibility, in agreement with our $D = 5$ behavior. Evidently, our sample is very close to the boundary between the two behaviors.

It is worthwhile to study both solvent types. The next step is to calculate the chemical potential curves, shown in Figure 6. The curves for water, $\mu_w$, are calculated via Sec. III D 2, and curves for solvent, $\mu_s$, via Eq. (15). These curves provide a different way of understanding the distinction between $D = 4$ and $D = 5$. The curvature of $\mu_s$ at $\phi = 0$ tells whether a dilute mixture is thermodynamically possible. For $D = 4$, $\mu_w''(0) < 0$, so the $D = 4$ fit represents a good solvent in this limit. Conversely, for $D = 5$, $\mu_w''(0) \gg 0$, so the $D = 5$ fit represents a very poor solvent that does not mix at small, finite concentrations. The two models disagree so flagrantly in this limit because all the measurements were taken for concentrated mixtures. For clarity, each data point shown is the average over the three runs. The data are plotted twice; once for the dry polymer mass $M_p$, found by the $D = 4$ model and again for the $M_p$ found by the $D = 5$ model. The $\mu_w$ (dashed, $D = 4$ below, $D = 5$ above) are fit according to Sec. III D 2. The resulting $\mu_s$ curves (solid, $D = 4$ top, $D = 5$ bottom), used in the Scheutjens-Fleer calculation, agree up to a constant in the range where data are available but disagree elsewhere.

**IV. CALCULATION RESULTS**

Using the experimental parameters found in Sec. III, this section shows the resulting predictions of polymer/water concentration profiles within the lamellae using the Scheutjens-Fleer methodology. Section IV A gives the results for the $D = 5$ model, which did not lead to phase separation in the lamella profile prediction. Section IV B gives the results for the $D = 4$ model, which does lead to predictions of phase separation, with a dilute channel layer in the center of the lamella.

Evidently, in the model PVBTMA polymers studied here, solvation effects dominate over chain stretching. This is due to the fact that $(V_K/V_M)_{20°C} \gg 1$, the volume of a PVBTMA Kuhn segment is much larger than the volume of a water molecule. This ratio appears as a prefactor in the definition of $\mu_K$ (Eq. (16)), which in turn is in the exponential in the segment Boltzmann factor (Eq. (6)). Therefore, the profiles are biased toward being locally flat. This flatness was observed in our calculated $\phi(\zeta)$ profiles. For a typical case, $\phi$ only varies about 0.01% from one Kuhn-length layer to the next. Another contributing factor to the flatness is the fact that the brushes on either side of the lamella overlap significantly compared to the length of a chain, giving the chains the opportunity to

![FIG. 5. Average free energy of the PVBTMA-water solution per lattice site, $A/Nk_BT$, for the Flory-Huggins fits (Eq. (19)) for polynomial degree 4, 5, and 6. The dashed line shows the mutual tangent line, which describes the phase separation.](image1)

![FIG. 6. Plots of $\mu_w$ and $\mu_s$ (Eq. (15)) for the $D = 4$ fit and $D = 5$ fit. The DVS data, averaged across runs, are plotted as vertical ticks. The data are plotted once for the $M_p$, found by the $D = 4$ model and again for the $M_p$ found by the $D = 5$ model. The $\mu_w$ (dashed, $D = 4$ below, $D = 5$ above) are fit according to Sec. III D 2. The resulting $\mu_s$ curves (solid, $D = 4$ top, $D = 5$ bottom), used in the Scheutjens-Fleer calculation, agree up to a constant in the range where data are available but disagree elsewhere.](image2)
thoroughly cover the entire lamella. For example, of 10,000 chain configurations sampled at $a = 0.58$, 31.5% reached or crossed the center plane. The exception to the flatness occurs when the PVBTMA and water solution are expected to phase separate, as is the case for the $D = 4$ model.

In order to find all possible self-consistent solutions for each brush spacing, $m$, several starting profiles were attempted including a flat profile, and water channels of varying widths. Also, to discourage the algorithm from skipping over solutions, the step size was limited so that for any $\xi$, the change in $\phi(\xi)$ per iteration was less than 0.02. Some initial guesses never converge; these runs were discarded. For calculations that did converge to a self-consistent profile, we calculated the resulting water activity (Eq. (12)) and free energy per chain (Eq. (17)). These points are connected to show how the membrane swells versus water activity.

In each section, the predicted swollen brush width, $w$, is compared to the experiment. These values are inferred by combining the SAXS $d$-spacing with the DVS mass uptake, similar to the calculation of $\sigma$ in Sec. III C. The brush width, $w$, is the average width of a hydrophilic block, i.e., half the total hydrophilic lamellar spacing. It equals the swollen volume of a hydrophilic block, $v_2 + v_w$, divided by the average interfacial area per chain, $\sigma^{-1}$.

\[
 w = (v_2 + v_w) \sigma = \frac{(v_2 + v_w) d}{2 (v_1 + v_2 + v_w)}.
\]

This forms an independent cross-check, since two independent experimental measurements, $v_w$ and $d$, were collected at each water activity, but only one combination of them, $\sigma$, was previously used. Between $a = 0.03$ and $a = 0.90$, the brush width increases from 8.88 nm to 11.60 nm, or $4.78\ell_K$ to $6.24\ell_K$. This does not afford much resolution in the calculation, which is limited to increments of $0.5\ell_K$ in brush width. Therefore, in Secs. IV A and IV B, we quadrupled the chain molecular weight to $n = 88$. The reported lengths are scaled by $1/a$, so that the effective spatial resolution is $\ell_K/4 = 0.465$ nm. This is an established way to scale the resolution, but it has the side effect of giving the chains more stretching entropy. This is moot in our case since the scaled system is still within the regime of negligible stretching energy, so it does not significantly affect the results.

Both models are in quantitative agreement with the experiment, with the $D = 5$ model giving a slightly better prediction of swelling. No experimental evidence of water channels is seen, but their presence cannot be ruled out. Sources of experimental error include polydispersity in the polymers used, inhomogeneity within the membrane, uncertainty in humidity measurements especially at high humidity, assumptions made in extracting the swelling width such as ideal mixing, and using the chlorinated and brominated form of the polymer interchangeably (see Appendix A). Sources of error in the calculation include ignoring effects smaller than a Kuhn length, ignoring dipole effect due to non-uniformity (see Appendix E), uncertainty in fitting the DVS data, not having a measurement of the characteristic ratio, and ignoring interaction with the hydrophobic layers other than setting the grafting density.

### A. Without phase separation ($D = 5$)

This section uses the best overall fit to the DVS data, i.e., Eq. (19) with $D = 5$ and the parameters given by Table II. This model predicts that untethered PVBTMA is not soluble in water below $\phi = 0.722$ (Figure 5). Figure 7(a) shows the shapes of the profiles as the water activity is increased from 0 to 1. From dry to water activity $a = 0.759$, the lamella swells gradually, expanding in width by only 3 Kuhn lengths (there are only four profiles spanning this humidity range). Above $a = 0.759$, the lamella swells linearly with increasing $a$, but at a faster rate. The two swelling rates can be seen in Figure 7(b), where each circle point corresponds to one profile, with matching color. The transition point, $a = 0.759$, between the two swelling rates corresponds to the minimum in grafting density (Figure 3(b)). Below this point, $\sigma$ decreases with $a$, tending to counteract the swelling. Above it, $\sigma$ increases with $a$, enhancing the swelling. The calculation predicts a water channel forms at $w = 11.6$ nm and $a = 1.0$. After that point, the lamella swells solely from an increase in the water channel width; the profile maintains a fixed shape. Figure 8 shows that the water channel width grows 1-to-1 with lamella width.

Compared to the swelling inferred from combined SAXS and DVS in Eq. (21) (Figure 7(b) black squares), the predicted swelling agrees to within about 10%, which was roughly the accuracy of the parameters fed into the calculation. The prediction somewhat over-predicts swelling for $a < 0.759$ and under-predicts the swelling for $a > 0.759$. Nevertheless, the good agreement supports the case that water is distributed very uniformly within the lamellae.

The model predicts that the free energy bottoms out at a lamellar width of 23.3 nm. Self-consistent profiles continue to be possible above this width, but as Figure 9 shows, the free energy does not vary beyond this point. The profiles remain the same shape, just with added water in the center. Therefore, any swelling beyond this point is due to forces not considered by our calculation. In addition to the swelling data in Figure 7(b), another measurement was performed at $a = 1.0$. We soaked the membrane in water for 24 h and reports the following changes compared to the dry membrane: the $d$-spacing increases from 41.9 nm to 68.8 nm, while the mass uptake divided by the dry mass, $(M - M_p) / M_p$, was 1.728. Using Eq. (21) with $v_1$ and $v_2$ from Sec. III C and the block molecular weights from Sec. III A, this results in a maximum hydrophilic lamellar width, $2w$, equal to 56.3 nm. Therefore, the $D = 5$ model says that water channels should have been present in the soaked membrane, and that forces external to the PVBTMA layers caused the membrane to swell to that point.

Overall, the predictions match the experimental cross checks to within about 10%–20% and capture the qualitative features of the data.

### B. With phase separation ($D = 4$)

This section uses the best fit generalized Flory-Huggins equation of state (19) with $D = 4$, with the parameters given in Table II. This model predicts that untethered PVBTMA has a solubility gap in water between concentrations $\phi = 0.230$ and $\phi = 0.813$ (Figure 5).
Figure 10(a) shows that water channels form when the concentration drops to the critical threshold of $\phi = 0.813$. In this flat-profile regime, the tethering does not affect the composition of the two phases, it just ensures that all the dilute phase is centered together in the lamella. Any other arrangement of the two phases would necessitate more phase boundaries or a breaking of the mirror symmetry, so it is the chain entropy that sets this geometry. The actual compositions of the two phases vary slightly in the calculation due to discretization error, since the water channel width is artificially forced to take an integer multiple of the Kuhn length.

Figure 10(b) shows that the swelling predictions for $D = 4$ agree with some of the cross checks but not all. From 0% through 72.3% water activity, the predicted swelling according to this model follows quite closely to that of the $D = 5$ model, and both are in good agreement with the SAXS+DVS cross-check. These are the prediction points that correspond to flat profiles.

Above $w = 10.7$ nm, all the predicted profiles are water channel shapes. They fall within a narrow range of water activity, 0.709–0.730, until the brush width is nearly as wide as fully stretched chains. This narrow range of $a$ corresponds...
FIG. 9. Total Helmholtz free energy of the system, $A$, per chain compared to pure water (Eq. (17)) versus the PVBTMA lamella width. The free energy drops as the swelling increases, up to 23.3 nm. Additional swelling occurs solely from water channel expansion, which costs no free energy in our model, and the curve bottoms out. This suggests the actual water channel width is set by forces not considered in the calculation.

FIG. 10. (a) Profiles for the $D = 4$ DVS fit. See Figure 7 caption for details. The profiles are flat, except that the solubility gap causes a water channel to form for $w > 9.6$ nm between concentrations $\phi = 0.230$ and $\phi = 0.813$. There is slight jitter around these values since the water channel width must be an integer Kuhn length. The water channel ceases when all the polymer has moved to the dilute phase. Finally, the chains are fully stretched, leaving pure water in the center. (b) Water activity versus swelled brush width. The calculation agrees with experiment up until the point where water channels form, at $a = 72.3\%$. The profiles maintain this water activity until the water channel ceases.

to the narrow range in $\phi$ at the center plane (Eq. (12)). To maintain a constant $\phi$ at the center, polymer must move from the concentrated phase to the dilute phase as the lamella widens, increasing the channel width. It does by increasing the channel width by $1.38\ell_K$ on average for each increase in $\ell_K$ of the lamellar width (Figure 11(a)). As mentioned, this introduces some frustration at small channel widths due to discretization (Figure 11(b)).

The free energy, $A$, as a function of water activity (Figure 12) is qualitatively similar to the $D = 5$ model, except that the point at which it bottoms out corresponds to a narrower swelled width, $2w = 19.2$ nm. Beyond this point, at little to no free energy cost, the system can move between any of the water channel widths. This could cause a swelling spike at $a = 0.78$, but no effect like this was observed. Conversely, for the water channel to swell larger than one Kuhn length, it would require expansion forces not considered in our calculation.

V. DISCUSSION

A. Overall assessment

Our motivation for this research was to understand the water distribution in ion-conducting polymer membranes and hopefully find ways to improve them. The goal was to advance our understanding by using experimental measurements in places where previous theoretical studies used estimates and analytics. This approach could have resulted in simply a
FIG. 11. (a) Water channel width versus total PVBTMA lamella width for the profiles in Figure 10. A water channel forms once the PVBTMA layer width, $2w$, has grown beyond 19.2 nm. The water channel width grows as $1.38(2w) - 26.5$ nm (dashed line). (b) Zoomed in around the transition region, showing discretization artifacts.

pragmatic advancement—a numerical validation of polymer brush theory. Instead, our results suggest a paradigm shift in the understanding of conducting blocks used in fuel cell membranes. We started with a polymer-brush mindset that the principal driver in the energy balance is between stretching and wetting. Using realistic parameters led to a different balance. First, the large Kuhn segment volume divided by solvent volume puts the polymer in a regime where stretching forces are secondary. Many of the predicted water distributions are surprisingly banal: uniform profiles with concentration determined via $\sigma(a)$, the areal density as a function of water activity. Therefore, measuring $\sigma(a)$ was critical in matching the results to experiment. Second, the realistic mixing effects between PVBTMA and water led to the surprising prediction of water channels. We showed that small changes in the mixing can lead to qualitative differences in the water concentration of the channel and to qualitatively different behavior in how the channels form and swell. There is limited evidence supporting

FIG. 12. Free energy per chain in units of $k_B T$ vs. PVBTMA lamellar width for the $D = 4$ model. See Figure 9 for details. After a water channel forms, the free energy and water activity remain nearly constant as the lamella swells. This suggests the water channel width is set by forces not considered in this calculation.
the existence of these channels. Given their potential for increasing ion conductivity in membranes, more study on the issue is warranted.

B. Parameters to the model

One of the advances of this research was to incorporate rigorous experimental measurements for the model parameters where simplifications had been used in similar previous models. One of the most important parameters, and the one we determined most carefully, was the water affinity of untethered PVBTMA. The measurements were performed to high precision using a sophisticated DVS apparatus and repeated three times. This was important, since we needed to extrapolate to more dilute concentrations than could be measured. It turned out that the extrapolation function we used, a generalization Flory-Huggins equation to polynomial degree $D$, was a good choice. The canonical Flory-Huggins equation, with $D = 2$, did not fit the data well, but the higher order equations did converge to the data. By eye, the $D = 4$ model fit well and we could have been forgiven for stopping there. By going to higher order, and using the AICc to compare models, we were able to rigorously show that the $D = 5$ model is the best fit. This turned out to be worthwhile, since the small difference in the $D = 4$ and $D = 5$ fitting resulted in a big difference in the extrapolation, and qualitative differences in the results. It should be noted that there is nothing inherently special about the numbers $D = 4$ or $D = 5$. These were simply the result of fitting our particular polyelectrolyte, and happened to illustrate two different phase behaviors. The AICc gave a quantitative understanding of the quality of fit, and we would recommend this methodology for fitting water uptake data. We would also recommend augmenting the water uptake data with additional data in the dilute regime, using light scattering, for example.

This paper rigorously derives the Kuhn segment chemical potential, $\mu_K(\phi)$, from the chemical potential of a water molecule in the system, $\mu_w(\phi)$, taking into account the osmotic pressure. In the past, many practitioners of the Scheutjens-Fleer method have used ad hoc, idealized forms for $\mu_K(\phi)$ that did not require this derivation. The canonical Flory-Huggins equation with its symmetric interaction term, $\chi N_p N_w$, cannot capture the wetting behavior of PVBTMA and similar complex polyelectrolytes used in fuel cells.

The Kuhn length is often thought of as the natural spatial resolution of a polymer, but it also determines the linear entropy density. It was this second role that was most important in the water profile calculations. A large Kuhn-segment-to-solvent-molecule volume ratio emphasizes mixing entropy over conformational entropy of the polymer. In this limit, brush effects are muted, and water tends to be homogeneously distributed. In contrast, studies of polymer brushes have assumed solvent molecules similar in size to a monomer. Some authors even make the mistake of equating a Kuhn segment with a monomer, which vastly overemphasizes conformational entropy. Although we did not measure the Kuhn length for PVBTMA, we can safely assume that it is in this limit of large Kuhn segments. Changing the value by $\pm 20\%$ turns out to have a negligible effect on the water distribution profiles. This choice in Kuhn length implies there are only 22 segments per PVBTMA block, leading to unacceptably low spatial resolution in the water profiles. Using a smaller Kuhn length to improve resolution would also alter the physics of the model. Instead, we scaled up the molecular weight of the block to 88 segments, and scaled down the reported lengths by $1/4$. This is equivalent to scaling the Kuhn length only in situations where it is used for spatial resolution, and not in computing the volume ratio. We checked that scaling in this way does not affect the profile shapes except to add more points.

In situations where the water profile is flat, the factor that governs the PVBTMA lamella width is the chain areal density, $\sigma^{-1}$. Even though $\sigma$ only varies by about 10%, it was precisely this effect that enabled the predictions to quantitatively match experiment, especially at low water activity. This effect is usually not included in polymer brush swelling calculations. We included the effect as an afterthought and had to use experimental measurements of $\sigma$, diminishing the predictive power of the model. One place for improvement would be to explicitly include the hydrophobic block in the calculation via separate $\phi_{\text{PMB}}$ and $\phi_{\text{PVBTMA}}$ functions. Two interaction potentials, $\mu_{\text{PMB}}$ and $\mu_{\text{PVBTMA}}$ would be needed, and each would depend on both $\phi_{\text{PMB}}$ and $\phi_{\text{PVBTMA}}$. The original calculation circumvented this complication by studying an A-B block copolymer with homopolymer A as the solvent. It may be worth the effort in order to understand the nonlinear behavior of $\sigma$.

Another way to improve our model would be to refine the assumption of ideal mixing, used in several places of the calculation. Ideal mixing is likely to break down at high polymer concentration, and could affect the conversion from the raw DVS data to the $\mu_K(\phi)$ used in the calculation. A more realistic mixing behavior could be incorporated using experimental measurements or MD simulations. Similarly, we made assumptions in several places about the density of various forms of PVBTMA: chloride vs. bromide forms, copolymer vs. untethered, and different percentages of monomers with a charged group. It would have been less ambiguous to measure these quantities directly.

C. Assessment of the results

The results differ from expectations in two ways: the predicted water distributions are strikingly uniform, except in cases of phase coexistence where abrupt water channels form.

First, we discuss the flatness aspect. This prediction is surprisingly different from the broad, parabolic profiles seen in ideal polymer brushes. This is the prediction for which we have the best experimental cross-checks, with x-ray scattering and water uptake. The agreement is good, to within about 10%, which is about as much as we could hope for given the uncertainty in some of the parameters we used. Since the polymer brush stretching effect turns out to be negligible, the same result could be found without a mean-field calculation. That is, the water uptake for PVBTMA in the membrane is nearly the same as the water uptake for untethered PVBTMA. These two water uptake functions should be compared, if possible, before deciding whether to study the
water distribution in a polymer membrane. Their degree of dissimilarity is a proxy for the degree of inhomogeneity of water.

The second surprising prediction is that our hydrophilic polymer is not miscible at all concentrations. At certain water activity, there is a phase coexistence, and the phases arrange to form water channels. They are only predicted to form within a narrow range of water activity. If the transition occurs at $a = 1$, such as our $D = 5$ model predicts, it would be straightforward to achieve by soaking the membrane in water. If the transition occurs at lower water activity, such as our $D = 4$ model predicts, then it would require sweeping the water activity slowly while holding the temperature fixed. Experimentally measuring the water profile shape would involve sophisticated scattering techniques, such as small-angle neutron scattering or anomalous x-ray scattering.

D. Water channels and conductivity

For any given polyelectrolyte, there is a certain non-zero concentration that maximizes ion conductivity. If there is not enough water, the ions do not diffuse. If instead there is not enough polymer, then the concentration of ions suffers. Preliminary evidence suggests this value for PVBTMA may be around $\phi = 0.2$.\(^{67}\) Therefore, the way to maximize conductivity with respect to water uptake is to focus the water into pathways as close to the optimal concentration as possible, in this case $\phi = 0.2$. The results for the $D = 4$ and $D = 5$ models show that there are two possibilities for water channels. In the $D = 5$ model, the channel is practically pure water. The only way for a pure water channel to conduct significantly is by the phenomenon in Appendix E, where ions from the concentrated phase dissociate and leak into the water phase. In the $D = 4$ model, however, the channel is at a finite concentration of $\phi \approx 0.22$. For maximizing conductivity, a polyelectrolyte with thermodynamics similar to the $D = 4$ model is probably better. In the $D = 4$ results, the water channel forms when the water activity is very close to $a = 0.82$. The fact that this activity is less than 1 is desirable, since fuel cells become flooded at high water activity. The fact that the channels only form at a specific water activity, though, is not desirable; it is unclear whether a real system could be maintained within such a narrow water activity band. One possibility to increase the range of conditions over which channels form is to add an even more hydrophilic block to the ends of the PVBTMA chains.

VI. CONCLUSION

In this paper, we used established polymer-brush theory to predict the spatial distribution of absorbed water absorbed in the hydrophilic lamellae of a diblock membrane. Under most normal operating conditions, the calculations predict a nearly uniform distribution of water. Under certain conditions, however, water is preferentially absorbed near the center plane of the hydrophilic lamella, according to our calculation. These water stripes can form if and only if a mixture of untethered hydrophilic block would phase separate under the same conditions. That is, two mixtures at different concentration coexist in equilibrium (both with the same chemical potentials) at the surrounding humidity level. In that case, there is no solvation energy penalty for having non-uniformities in the polymer/water ratio, as long as the disparate regions are coexisting phases. The conformational entropy generated by tethering each hydrophilic chain to an interface serves as a tie-breaker between otherwise equivalent states.

Using the methods and results from this study, it should be possible to engineer polymers that form water channels more easily. If they can be made to be robust, it provides a way to concentrate water into conducting pathways. This arrangement maximizes ion diffusivity for a given amount of water and leaves the majority of the membrane relatively dry and strong. Water channels have recently been observed in proton-exchange membranes.\(^{25}\) The next step will be to generate them in alkaline anion exchange membranes and experiment with their properties. Hopefully, this work is a step toward cost-efficient, viable alternatives to batteries and combustion engines.

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APPENDIX A: BR/CL INTERCHANGEABILITY

The Scheutjens-Fleer method requires the solubility function of the isolated (untethered) hydrophilic block (Sec. III D). Due to differences in synthesizing the block copolymer versus the isolated hydrophilic block, the chloride form of the polymer was used for the DVS measurement whereas the bromide form was used in the membranes. The following argument shows that both forms should have nearly the same solubility when measured as water activity vs. volume fraction. Cohen \textit{et al.}\(^{68}\) performed very accurate water uptake experiments on droplets suspended by an electric field comparing NaCl versus NaBr. They report the data as a polynomial fit water activity vs. molality, or moles of substance per kilogram of water. To convert from molality to volume fraction, $\phi$, requires the condensed density of the substance, $\rho$, and the molar mass, $M$. For example,
bromine has a molar mass of $M_{\text{Br}} = 0.0799$ kg/mol and a condensed density of $\rho_{\text{Br}} = 3100$ kg/m$^3$. Ignoring the sodium for simplicity, a solution of volume $V$ has $V\phi_{\text{Br}}\rho_{\text{Br}}/M_{\text{Br}}$ moles of bromide, and $V(1-\phi_{\text{Br}})$ $\rho_w$ mass of water, where $\rho_w = 1000$ kg/m$^3$ is the density of water. Therefore,

$$\text{molarity} = \frac{\phi_{\text{Br}}\rho_{\text{Br}}/M_{\text{Br}}}{(1-\phi_{\text{Br}})} \rho_w.$$ 

The analogous relation applies to chlorine using a molar mass of $M_{\text{Cl}} = 0.0355$ kg/mol and a condensed density of $\rho_{\text{Cl}} = 1560$ kg/m$^3$. Figure 13 shows plots of the functional form measured by Cohen et al. converted to volume fraction. As the plot shows, the two functions agree to within 6% of each other up to the volume fraction of bromide found in pure PVBTMA $[\text{Br}^{-}]$. We expect the equations of state for PVBTMA $[\text{Br}^{-}]$ and PVBTMA $[\text{Cl}^{-}]$ to be in similar agreement. Therefore, it should be acceptable that we use the chlorinated form of the hydrophilic block in Sec. III D to make predictions about the brominated membrane.

**APPENDIX B: SCHEUTJENS-FLEER ILLUSTRATION**

Equations (9)–(11) use the “Scheutjens-Fleer trick” to streamline the calculation on the local concentration ($\phi(\zeta)$). Here, we demonstrate the efficiency gained in a small-scale example. Table III illustrates a direct calculation over all chain conformations versus Table IV which uses the Scheutjens-Fleer trick. Both methods are applied to a 3-segment chain, using the segment Boltzmann factors, $T_\zeta \equiv \exp[-\mu(\zeta)/k_BT]$. Naively, to compute the Boltzmann-weighted average chain profile, $\langle \phi(\zeta) \rangle$, one would follow Table III, enumerating all the conformations, calculating the relative probability, $R$, of each conformation, and computing the profile using Eq. (7). This is computationally expensive as there are on the order of $3^n$ conformations. The Scheutjens-Fleer trick is much more efficient. It involves an ostensibly similar calculation, shown in Table IV, performed once for $i$ increasing and again for $i$ decreasing. The difference in workload is that each term is immediately computed to a numerical value, so the algebraic complexity does not grow. Even though information is lost by collapsing the terms, the trick is that the profile $\langle \phi(\zeta) \rangle$ can still be determined by combining both versions of the calculation in a clever way, shown in Eq. (8).

**APPENDIX C: HALF-INTEGER MIRROR BOUNDARY**

Section II C describes how to find the self-consistent polymer volume fraction profile for a total hydrophilic width of $2m$. To get more resolution, the method can also accommodate odd-integer spacings by shifting the mirror boundary condition from the end of the $m$th Kuhn cell to the center. This modifies Eqs. (9) and (10) for $\zeta = m$ to

$$G_{i-\zeta}^{1-m} = \exp[-u_K(m)] \{2G_{i-\zeta}^{1-m-1} + 4G_{i-\zeta}^{1-m}\}$$

and

$$G_{(i-1)-m}^{m-\zeta} = \exp[-u_K(m)] \{2G_{i-n}^{m-\zeta} + 4G_{i-n}^{m-\zeta}\},$$

respectively. Cell $m$ is then divided between the left and right sides of the lamella.
APPENDIX D: CHECK OF AREAL-DENSITY SCALING

The grafting densities inferred in Table I can be cross-checked by the following rough analysis. The grafting density varies in such a way as to minimize the total energy of the system. The surface tension of the hydrophilic/hydrophobic interface, $\gamma$, acts to minimize the surface area $\sigma^{-1}$. The surface energy per chain is given by $U_S = \gamma \sigma^{-1}$. We neglect any dependence of $\gamma$ on water uptake. The entropy due to stretching of each block acts to minimize the width of its respective lamella. Let $U_1$ and $U_2$ be the stretching energy per chain of the hydrophobic block and hydrophilic block, respectively. Then,

$$U_i = \frac{k_B T}{2} (\eta v_i \sigma)^2,$$

where the subscript $i$ takes value 1 for the hydrophobic block and 2 for the hydrophilic block. Here, we have assumed that the average end-to-end distance of the chains can be written $(\eta v_i \sigma)$, where $(v_i \sigma)$ is the lamellar width and $\eta$ is a fixed proportionality constant. The total free energy is $F = U_S + U_1 + U_2$. The condition for a minimum in free energy is $F'(\sigma) = 0$. This results in a scaling law,

$$\sigma \propto \left( \langle r^2 \rangle_{0,2} v_1^2 + \langle r^2 \rangle_{0,1} (v_2 + v_w)^2 \right)^{-1/3}. \quad (D1)$$

The number of repeat units per chain is $n_{b,1} = 358$ and $n_{b,2} = 160$ for PMB and PVBTMA, respectively. The characteristic ratio for PMB is $C_{\infty,1} = 7.2^{50}$ and for PVBTMA is $C_{\infty,2} \approx 9.85$ as discussed in Sec. III B. The bond length, $\ell_b$, should be the same for each. Taken together, this gives

$$\frac{\langle r^2 \rangle_{0,2}}{\langle r^2 \rangle_{0,1}} = \frac{C_{\infty,2} n_{b,2} \ell_b^2}{C_{\infty,1} n_{b,1} \ell_b^2} = 0.61.$$

Figure 14 shows how the scaling law fits the measured grafting densities from Table I. The scaling law follows the downward trend in the points, until the last point which ticks up. This trend has also been measured in other experiments.$^{54,55}$ It is likely due to another force becoming relevant, possibly due to geometric constraints preventing the lamellae from swelling laterally past a certain point.

APPENDIX E: HYDROXIDE DISTRIBUTION IN WATER CHANNEL

There are two reasons to study the electric field induced by a non-uniform polyelectrolyte/water distribution. The first is to perform a self-consistency check: the electric field was ignored in the methodology of this study, and we need to check that this is a reasonable assumption. The second is to quantify the effect in water channels of bonus ions—hydroxide ions that dissociate from the concentrated phase and diffuse into the dilute phase, adding to the conductivity of the channel.

Electric fields develop due to a water channel, but these turn out to be small. In Sec. IV, we inferred the free energy cost of a polymer-water mixture from the water uptake experiment (DVS) of the untethered hydrophilic block, implicitly assuming that the properties are equivalent. This assumption could break down, however, if the free ion distribution changes in response to the tethering forces. Then, an electrostatic gradient could develop that was not accounted for in the analysis. Here, we show that the effect is small enough to ignore when calculating the polymer profiles. The free ions are not tied down to their counter-ions on the polyelectrolyte and are free to thermally diffuse. Since the polyelectrolyte and the free ions are subject to different constraints, they follow different spatial distributions and the resulting segregation produces electric fields. To place a bound on this effect, we analyze the worst-case scenario, where the
polyelectrolyte distribution jumps discontinuously from one value to another (Figure 15). The free ion distribution is then determined by the Boltzmann weighting

$$\rho_-(z) = \rho_1 \exp \left[ -\frac{q\Psi(z)}{k_BT} \right].$$

(E1)

where $\Psi(z)$ is the electric potential at point $z$, $q$ is the elementary charge, and $\rho_1$ is the (negative) charge density at the interface due to OH$^-$ (this sets the gauge by requiring $\Psi = 0$ at the interface). Plugging this expression into the Poisson-Boltzmann equation gives the Poisson-Boltzmann equation

$$-\frac{\partial}{\partial z} [\epsilon(z) \frac{\partial \Psi}{\partial z}] = \rho_+(z) + \rho_1 \exp \left[ -\frac{q\Psi(z)}{k_BT} \right].$$

(E2)

Here, $\epsilon(z)$ is the background electric permittivity. To avoid the problem of a discontinuous function for $\epsilon(z)$, here we use a uniform permittivity, $\epsilon(z) = 40\epsilon_0$, an estimate that is roughly midway between that for hydrocarbons and water. To find $\rho_+(z)$, we first estimate that the charge density for pure PVBTMA [Cl$^-$] is roughly 3.8 $q$ nm$^{-3}$. This value is combined with a contrived “worst-case” step function to give the fixed charge density distribution. Because the Poisson-Boltzmann equation is highly non-linear and very unstable numerically, we isolate a 2 nm region, $z \in [-2 nm, 2 nm]$ containing the step function

$$\rho_+(z) = 3.8q \times \begin{cases} 0.2 & z > 0 \\ 0.8 & z < 0 \end{cases}.$$  

The differential equation is order 2 and so requires two boundary conditions. We set $\Psi(0)$ such that $\rho_- (0) = \rho_+(0)$, and iterate choices of $\Psi'(\approx 2 nm)$ until $\rho_-(2 nm) = \rho_+(2 nm)$. Note that if $\Psi'(\approx 2 nm)$ and $\Psi'(\approx 2 nm)$ both vanish, then the solution outside the box will satisfy $\rho_- (z) = \rho_+(z)$. Such would be the case if the calculation domain was large enough to contain the transient due to the sharp step function. This turns out to be the case. It turns out that $\Psi'(\approx 2 nm) \approx -3.62 \times 10^{-8}$ V/nm and $\Psi'(\approx 2 nm) \approx -2.31 \times 10^{-5}$ V/nm, which are both quite small for the scales involved in the problem.

The resulting hydroxide profile is a smoothed-out version of the polyelectrolyte profile. The free charges pull on the fixed charges, which will tend to smooth out the discontinuity in the polyelectrolyte distribution, but this happens on a length scale that turns out to be smaller than the Kuhn length. The characteristic width of the mismatch between the two distributions is given by the Debye length,

$$\Lambda_D = \left( \frac{e k_B T}{q \rho} \right)^{1/2} \approx 0.2 \text{ nm},$$

which is about ten times smaller than the Kuhn length. Therefore, it was a reasonable assumption to ignore the smeared charge distribution in calculating the polymer distribution.

Figure 15 illustrates that there are “bonus” hydroxide ions in the water channel due to the dissociation effect. In the case shown, there are 1.057 hydroxide ions per counterion in the channel, only a modest increase. We conclude that the bonus ion effect is small for water channels with a finite concentration. The same effect for a pure-water channel, however, could have a relatively large effect on conductivity.

The scaling behavior for the density of bonus ions can be found analytically if the water channel is much wider than the Debye length. In that case, we only have to solve for the charge distribution surrounding one of the water channel boundaries. As before, we focus on one of the water channel boundaries and model the fixed charge distribution there as a step function. Let $\rho_{avg}$ be the average fixed charge distribution of the two phases, and let $\delta$ be a variable that parametrizes the step height. Then, the fixed charge distribution can be generalized as

$$\rho_+(z) = \begin{cases} (1 + \delta) \rho_{avg} & z \leq 0 \\ (1 - \delta) \rho_{avg} & z > 0 \end{cases}.$$  

(E3)

We used Mathematica to solve the linearized Poisson-Boltzmann equation, where the exponential in Eq. (E2) is replaced by its Taylor expansion to linear order. For simplification, we used a uniform, average electric permittivity, $\epsilon(z) = \epsilon_{avg}$. The resulting differential equation is

$$-\epsilon_{avg} \psi''(z) = \rho_+(z) - (1 + \delta) \rho_{avg} (1 + q \psi(z)/k_BT).$$

We define the boundary conditions on an interval, this time generalized as $[-L, L]$ for some $L \gg \Lambda_+$. Similar to before, the boundary conditions are $\psi(-L) = 0$ and $\psi'(L) = 0$. Using a shorthand for the Debye length of the concentrated phase, $L_+ \equiv \left[ \epsilon_{avg} k_B T / (1 + \rho_{avg} \rho_B) \right]^{1/2}$, the solution for the potential is

$$\psi(z) = -\frac{\delta k_B T}{(1 + \delta) q \cosh[L_+/z_0]} \left\{ \begin{array}{ll} \sinh[L_+/z_0] & z \leq 0 \\ \cosh[L_+/z_0] - \cosh[L_+/z_0] \cosh[L_+/z_0] & z > 0 \end{array} \right.$$  

The number of bonus ions divided by water channel interface area, $\Sigma_B$, is defined by the area between the $\rho_-$ curve and $\rho_+$ level for $z > 0$,

$$\Sigma_B = \int_0^L [\rho_-(z) - \rho_+(z)] \cdot dz.$$  

(E4)

Substituting in the solved form for $\psi(z)$ and integrating with Mathematica results in

$$\Sigma_B = \delta \frac{\sqrt{\epsilon_{avg} \rho_{avg} k_B T}}{(1 + \delta) q} \tanh \frac{4 (1 + \delta) L^2 q \rho_{avg}}{\epsilon_{avg} k_B T}.$$  

The tanh term is a transient that asymptotes to unity in the limit $L \gg \Lambda_+$. In that case, the bonus ion density is

$$\Sigma_B \approx \delta \frac{\sqrt{\epsilon_{avg} \rho_{avg} k_B T}}{(1 + \delta) q}.$$  

Thus, the bonus ion effect has an approximately linear dependence on $\delta$, the disparity in concentration between the two phases, and a $1/2$ power dependence on the average concentration.