The Role of Noise in Determining Selective Ionic Conduction Through Nano-Pores

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Abstract—The problem of predicting selective transport of ions through nano-pores from their structure in the biological and nano-technological systems is addressed. We use a molecular dynamics simulation to provide insight into the key physical parameters of nano-pores and develop a self-consistent analytic theory describing ionic conduction and selectivity through these devices. We analyse the ion’s dehydration and excess chemical potential, derive an expression for the conductivity of the nano-pore, and emphasize the role of fluctuations in its performance. The theory is verified by comparison of the predicted current-voltage characteristics with the molecular dynamics results and experimental data obtained for a graphene nano-pore and the KcsA biological channel.

Index Terms—Graphene nanopore, biological ion channel, selectivity, functionalisation, molecular dynamics, statistical theory, kinetic theory.

I. INTRODUCTION

The investigation of ionic transport through nano-pores is a vast and rapidly expanding field encompassing both the nano and bio-physics communities. These pores have radii on sub-nanometer length scales and can be as thin as a single layer of graphene; and yet their selectivity properties are difficult to predict and/or control. Interest in these pores is stimulated both by the wide range of applications including e.g. the desalination of water [1], and by the novel light they shed on understanding the function of biological ion channels [2]. In particular, the bio-inspired functionalisation of these nano-pores offers a challenging avenue for the composition of nano-scale devices with pre-defined properties. Another long-standing problem in the analysis of these devices is the development of an analytic theory to describe the trade off between their high selectivity and fast permeability [3]–[5].

In this work we aim to address these challenges by introducing a self-consistent approach to investigating the effect of hydration and fluctuations on the performance of nano-pores. We develop statistical and kinetic theories of their conductivity and selectivity. Theoretical predictions are compared with experimental and simulation data of conduction through a single-layer graphene pore and a biological channel KcsA.

II. ION PERMEATION THROUGH A PORE IN GRAPHENE

In a preliminary simulation study we have compared the properties of a neutral and a charged pore opened inside a single graphene sheet. The aim of these simulations is to highlight the most important factors affecting ion permeation and investigate the role of hydration shells. Both charged and neutral systems (left panel on Fig. 1) comprise a single fixed graphene layer with a pore hydrated on both sides by a 1M KCl solution. The minimal carbon-carbon distance in the pore is ~ 0.45nm, and so the effective pore radius accounting for the hydrophobicity is ~ 0.23nm. The dimension of the simulation box is 2.6×2.4×4nm, and following Ref. [6] the geometry of the pore was chosen to match the shape of one of the four carbonyl rings in the selectivity filter (SF) of KcsA channel. To create the charged system, four C atoms on the pore rim were assigned charges ~−0.5q (matching the oxygen of the carbonyl groups), and the electro-neutrality of the system was enforced by a balancing charge of +2.0q uniformly spread over all the remaining graphene atoms.

In order to obtain the potential of mean force (PMF) of the permeating potassium ion (K+) the umbrella sampling method was employed. In these simulations the axial coordinate z of a single K+ was harmonically restrained while the radial position was subject to a semi-harmonic wall at 0.45nm from the pore axis. We considered 61 axial windows from z = −1.5nm to 1.5nm with width of 0.05nm. Each window was run for 1.5ns with the first 0.5ns considered as equilibration and excluded from the analysis. Using the implementation by Grossfield [7] the weighted histogram analysis method (WHAM) [8] was used to calculate the PMF with a tolerance of 10−7.

Both the charged and neutral PMF are highlighted in Fig. 2 as red and blue respectively. The neutral pore produces a barrier of 7.31kT, whilst the charged pore yields a barrier
of 2.81kT agreeing with a similar set up as reported by Sahu et al. [9], and He et al. [6] for a 4CO system. This suggests that the height of the barrier is primarily determined by the charge of the pore and not by the specific geometry or placement of the functional groups. The fact that this barrier is much smaller than the one computed for the neutral system confirms that the presence of negative charges in the pore favours cation permeation. The PMF in Fig. 2 also shows the existence of two minima on either side of the graphene monolayer, centered at \( z = -0.5 \text{nm} \) and +0.5nm. The two free energy minima arise because in these locations the \( K^+ \) is close enough to the pore to feel the electrostatic attraction of the charges but it is still not in the gap region between the water phase and the graphene layer, so it is still not significantly dehydrated.

The electrostatic component \( G_C \) of the charged PMF is related to the Coulomb potential between \( n_f \) negative pore charges, and the ion \( q \) (see right panel in Fig. 2), which decreases as the ion approaches the pore, taking the average value \( \sim 9 \text{ inside} \). This indicates that water in the pore should not be treated as in the bulk, and can be explained by noting that, despite the minor dehydration (see subsequent discussion and the right panel of Fig. 1), the majority of the charge-screening water molecules in the shells reside outside of the pore.

To investigate the role of hydration in more details, we have computed the number of water molecules in each hydration shell surrounding a restrained \( K^+ \) ion in the windows of the umbrella sampling simulations. As shown in Fig. 3, both the first and the second hydration shell lose water upon crossing the pore but the depletion is more significant for the second shell, in agreement with the results of Sahu et al. [9]. This occurs because, when the \( K^+ \) ion crosses the pore, it tends to stay close to the charged carbon atoms establishing direct salt bridges, thus minimising the available space for coordinating waters. On the other hand, the water molecules of the second shell lying further outside the pore are still attracted electrostatically to the charged rim as well as the ion, and thus still remain in the neighborhood of the ion. For analytic calculations one needs to estimate the dehydration energy for each shell. To do so, Zwolak and coworkers [10] proposed to couple the dehydration energy and the cost for its change with shell’s surface area. Thus, the energy barrier becomes a function of the pore radius \( R_p \): \( \Delta U(R_p) = \sum_j (f_j(R_p) - 1)U_j^0 \). Here the sum runs over all \( j \) shells, and the dehydration fraction \( f_j \) is the remaining portion of shell \( j \) left upon permeation through a pore of radius \( R_p \). The energy \( U_j^0 \) stored in each hydration shell is

\[
U_j^0 = \frac{zq^2}{8\pi\varepsilon_0} \left( \frac{1}{R_{Ij}} - \frac{1}{R_{Oj}} \right) = \frac{zq^2}{8\pi\varepsilon_0} \left( \frac{1}{\varepsilon_p} - \frac{1}{\varepsilon_w} \right)
\]

where \( R_{Ij} \) and \( R_{Oj} \) are the outer and inner radii of the shell respectively, and \( \varepsilon_p \) is the permittivity of the membrane. The solid lines on Fig. 3 display this fraction for \( R_p = 0.23 \text{nm} \) and length \( L_c = 0.4 \text{nm} \). As long as the fraction is computed solely from the geometry, the dehydration profiles (1) are identical for both the neutral and charged systems. Although the estimation (1) lacks water-water and water-fixed charge interactions, one sees quite good agreement particularly for the neutral pore. The differences can be attributed in part due to the lack of these interactions, but also the influence of nearby ions and the cylindrical approximation for the pore geometry.

The above results justify the picture that \( K^+ \) ion outside the pore is not significantly dehydrated and hence the strong electrostatic attraction it faces produces the two free energy minima flanking the barrier in the PMF (Fig. 2). This interpretation is consistent with Fig. 3 where in positions \( \pm 0.5 \text{nm} \) the first hydration shell is practically intact while the second shell has only lost approximately 40% of its water molecules.

Thus, the results of our preliminary MD simulations indeed identify two driving forces for ion permeation: (i) the ion-pore electrostatic interactions and (ii) the dehydration cost the ion incurs when crossing the pore. This is in agreement with the seminal theory developed by Eisenman [3]. In the next section, we incorporate those energy contributions in statistical theory.

III. THEORY OF SELECTIVE CONDUCTION OF MOLECULAR PORE: ROLE OF FLUCTUATIONS

The preceding discussion demonstrates that, to take explicit account of dehydration and strong interaction between conducting and fixed ions at the pore, a self consistent theory
has to be defined at the molecular level. We note further that under typical experimental conditions the thermal de Broglie wavelength $\Lambda$, of ions of $i$-th type in the solution is of the order of $\sim 0.03 \text{nm}$. This is much less than the average distance $((V/n_i)^{1/3}$) between ions in the bulk and even in narrow pores and so the system is classical and the ions obey Boltzmann statistics. Therefore the physical properties of the whole system, comprising a pore and two bulk reservoirs, are determined by the canonical ensemble assuming that it conserves total volume $V$, temperature $T$ and the total number of particles of each species $N_i$. The total energy of this ensemble has the form [11]–[13]

$$E\{n_j\}, n_f) = E_0 + (N_w - n_w) \mu^c_w + n_w \mu^e_w - kT \ln \left( \frac{K^i_n}{n_w \prod_i \frac{1}{n_i}} \right) + \mathcal{E}\{n_j\}, n_f).$$

Here, $\{n_j\}$ is a set of ions in the pore, $E_0 = TS - pV$ is the thermodynamic part of system’s energy for a given entropy $S$ and pressure $p$. The next two terms in the first row explicitly account for the energy changes due to the transfer of the water molecules between bathing solutions and the pore.

The second row describes the energy required to add $n_i$ ions of the $i$-th type to the pore. In the last row the first term accounts for permutations of ions of each type and water molecules, while $E\{n_j\}, n_f)$ corresponds to ion-ion and ion-fixed charge interactions inside the pore. The electrochemical potential of the ions is $\mu^\text{b,c}_i = \mu^\text{b,c}_0 + \varphi \mu^\text{b,c}_i$, where $\mu^\text{b,c}_0$ represents its ideal part and $\mu^\text{b,c}_i = \mu^\text{b,c}_0 + q \varphi \mu^\text{b,c}_i$ consists of the electric $\varphi \mu^\text{b,c}_i$ and excess chemical $\mu^\text{b,c}_0$ potentials in the bulk (b) or in the pore (c). In what follows we neglect the difference in the thermal wavelength and the internal degrees of freedom between the bulk and the pore, and the effect of electric field on the motion of water molecules because $z_w \sim 0$.

The single-file conduction of ions through the pore with a finite number of binding sites $K$ defines the set of energy states of the pore and its total occupancy [12]–[17]. Observing that transitions of ions from the bulk to the pore correspond to small fluctuations in the total energy of the system [11] and canceling out constant terms in the energy Eq. (2) we obtain [12], [13] the free energy of such transitions

$$G\{n_j\}, n_f) = E\{n_j\}, n_f) + \sum_i n_i (\mu^\text{c}_i - \mu^\text{c}_0) - kT \ln \left( \frac{K^i_n}{n_w \prod_i \frac{1}{n_i}} \right).$$

The corresponding grand-canonical ensemble reads

$$P\{n_j\}, n_f) = \frac{1}{Z n_w \prod_i \frac{1}{n_i}} \exp \left( \sum_i n_i (\mu^\text{c}_i - \mu^\text{c}_0) - E\{n_j\}, n_f) \right).$$

where the grand partition function $Z$ is defined from the normalisation condition $\sum_{\{n_j\}} P\{n_j\}, n_f) = 1$, and yields the grand potential

$$\Omega = -kT \ln Z.$$

Thus we arrive at a statistical model of the pore that takes explicit account of dehydration (via $\mu^\text{c}_i$) and strong interaction (via $E$) between ions inside the pore.

In order to estimate $\mu^\text{c}_i$ and $\mathcal{E}$, and to introduce their analytic approximations, we notice that the probabilities $P\{n_j\}, n_f)$ for a given configuration of ions $n_i$ in the pore are related to the binding probabilities $\mathcal{B}\{n_j\}, n_f) \sim P\{n_j\}, n_f)/P\{0\}, n_f)$ introduced by Roux [18], [19] leading to

$$e^{-\left( \sum_i n_i (\mu^\text{c}_i - \mu^\text{c}_0) + \mathcal{E}\{n_j\}, n_f) \right) \sim \int e^{-\mathcal{W}(r_1, ..., r_n)} dr_0^n}$$

with $\mathcal{W}(r_1, ..., r_n)$ standing for the $n$-ion PMF of the pore and the integration runs over the pore volume. The analysis in Sec. II shows that the main contributions into the $n$-ion PMF are due to dehydration Eq. (1) [10] and to the local interactions between the mobile ion and surrounding charges. The latter can be approximated by electrostatic interactions with effective dielectric permittivity of the pore.

In addition, the global ion-ion interaction in a charged water-filled pore can be approximated analytically [15], [17], [20] using the notion of the charging energy of the pore familiar from the Coulomb blockade theory of quantum dots, see e.g. [21], [22]. The resulting expression is

$$\mathcal{E}\{n_j\}, n_f) = \frac{q^2 L_c}{8 \pi \epsilon_0 \epsilon_r R_c^2} \left( n_f + \sum_i n_i \right)^2$$

where the prefactor, can be denoted by $L_c = q^2/2C_s$ with the pore capacitance being $C_s$.

IV. SELECTIVITY VS CONDUCTIVITY TRADE-OFF

The current density $j_i$ through the pore can be written in general form [19], [23]–[25] as $j_i = -\frac{Z}{q} \nabla \mu_i$, where the conductivity $\sigma$ is defined using the generalised Einstein relation [12], [26], [27]

$$\sigma_i = q^2 D_i \frac{\partial c_i}{\partial n_i} \propto \langle \Delta n_i^2 \rangle = -kT \left( \frac{\partial^2 \Omega}{\partial n_i} \right)_{T,V}$$

where $c_i = \langle n_i \rangle / V_c$ is the concentration, $D_i$ is the diffusivity and $n_i$ is the chemical potential in the pore with volume $V_c$. The current is directly proportional to the fluctuations in the number of ions and, accordingly, current is largest when the rate of these fluctuations is maximal.

In the particular case of a three level system $\{0\}, \{X\}, \{Y\}$, where the ground state 0 corresponding to a given set $\{n_j\}$ of ions in the system is excited by adding an ion of type $X$ to $Y$ to the pore, the conductivity of each ion type is

$$\langle \Delta n_i^2 \rangle = e^{-\Delta G_i / kT}$$

where $\Delta G_i = G(n_i + 1, n_f) - G(n_i, n_f)$ stands for the free energy barriers and an analogous expression for $Y$ is obtained by exchanging symbols $X$ and $Y$.

It follows from equation (9) that maximum conductivity for a given type of ions (e.g. $X$) is achieved when

$$\Delta G_X \sim 0, \text{ and } \Delta G_X \ll \Delta G_Y,$$

1This charging parameter reflects the strength of interaction between the mobile ions and the pore charge.
hold simultaneously. The first relation represents the condition for barrier-less conduction. Given that transitions occur between \( N \) and \( N + 1 \) states, one recognises the knock-on mechanism. Eventually, the last condition embodies the strong selectivity regime. Thus, maximising the noise (variance) of pore occupancy, one derives the optimal conduction and selectivity properties. We note that it is surprising that exactly these conditions are found in Nature, namely in the biological KcsA channel.

Thus the optimal parameters of the nano-pore for selective conduction at nearly the diffusion rate can be now obtained analytically in the form

\[
n^*_j = \frac{C_s}{q^2} \left( \Delta \mu_X^* - kT \ln \frac{(n_X + 1)}{(K - n_X - n_Y)} \right) - \left( n_X + \frac{1}{2} \right),
\]

(10)

the optimal chemical potential difference (\( \Delta \mu_X^* = \mu_X^* - \mu_X^b \)) is given by

\[
\Delta \mu_X^* = \frac{1}{n_X} \mathcal{E}(n_X, n^*_j) + \frac{kT}{n_X} \ln \left[ \frac{(n_X)(K - n_X - n_Y)!}{(K - n_X - n_Y)!} \right].
\]

(11)

It is clear therefore that the pore allows for the unhindered (but still diffusion-limited) passage of these favoured ions, while selectively blocking the other species. The pore is tuned to satisfy this condition, resulting in the following energy barrier for an ion of the same valence

\[
\Delta \Delta G = (\mu_Y^* - \mu_Y^b) - (\mu_X^* - \mu_X^b) + kT \ln \frac{q^2}{e}\frac{C_X}{C_Y}
+ kT \ln \frac{n_Y + 1}{n_X + 1}.
\]

(12)

Provided that the free energy barrier for each type of ion to enter the pore Eq. (3) is a global function of the pore parameters, for ions of the same valence the difference between these barriers is determined mainly by the local interactions in the pore as envisaged by Eisenman [3]. This result is intuitively reasonable since on distances of a few ionic radii, the ions behave as point changes and cannot be distinguished. The above analysis is equally applicable for uncharged pores, different pore radii or lengths and ions of different valence.

Hence, we have shown that a generalised form (12) of Eisenman’s selectivity relation follows directly from the condition for the diffusion-limited conduction of \( K^+ \) ions. The corresponding conductivity of the favoured ion can now be written as

\[
\sigma_X = \frac{q^2 D_X e^{-\Delta \Delta G_X/kT}}{V_c kT \left(1 + e^{-\Delta \Delta G_X/kT}\right)^2},
\]

(13)

V. KINETIC THEORY

To investigate non-equilibrium responses we will introduce a self-consistent kinetic theory, describing permeation with rate \( \Gamma \) through the discrete state space identified in the previous section. The behavior can be analysed with a discrete set of master equations, where the transiting ion is denoted by \( n^*_j \),

\[
\hat{P}\{n_j\} = \sum_{i} \sum_{J} \sum_{n_j+n^*_j} P(n_j) \Gamma^{b,i}_{n_j,n^*_j} - P(\{J\}) \Gamma^{b,i}_{J,n_j},
\]

(14)

The left hand side of the equation delineates the temporal behavior and must equal zero in the steady-state regime. The sums run over species \( i \), both left and right bulks \( b \), and for clarity conditions \( * \). These conditions require that the state of the pore must remain within the state space and the transitions between these states be physically possible i.e. the number of ions cannot exceed the number of sites, and an ion can only exit if it is already occupying the pore. Formally this can be defined as \( 0 \leq J \leq K - n_u \) and when removing an ion \( n_j - n^*_j \neq \sum_{i \neq j} n_i \), where \( n^*_j \) denotes the sum over all ions not including the transiting \( i^n \) species. For notational clarity we drop the functional dependence on \( n_j \) although it remains present in the energy barrier.

The steady-state current can be defined as in standard practice from the balance of fluxes, such that Kirchoff’s laws are satisfied. It is defined such that its positivity defines ionic flow from left to right,

\[
I_i = \pm q \sum_{n_j=0}^{K-1-n_u} P(n_j) \Gamma^{b,i}_{n_j,n_j+n^*_j} - P(n_j+n^*_j) \Gamma^{b,i}_{n_j+n^*_j,n_j},
\]

(15)

The transition rates are defined such that at equilibrium the detailed balance conditions equate to the Boltzmann ratio with the equilibrium free energy barrier given by equation (3). The energy profile is assumed to take the form of a binding site whereby the particle must escape and therefore the influence from the potential drop is assumed to be solely inside the pore [28]. We note that this can be extended to include a chemical gradient if needed. The rates therefore take the following form,

\[
\Gamma^{b,i}_{n_j+n^*_j,n_j} = A_i e^{(\Delta G_i - z_i q (\phi^b - \phi^f))/kT},
\]

(16)

\[
\Gamma^{b,i}_{n_j,n_j+n^*_j} = A_i, \quad \text{where} \quad A_i = \frac{2D_i}{L_c^2 (1 + e^{-\Delta G_i/kT})},
\]

(17)

The equilibrium normalisation \( A_i \) must relate to the general energy profile of the system and in its linear response must reproduce Ohm’s and Fick’s laws for the favoured ion as calculated earlier in equation (13). It is important to note, however, that we do not include a transition state and nor do we explicitly track the motion of ions through sites. These developments will be considered in further publications.

VI. COMPARISON WITH EXPERIMENTAL DATA

A. Example 1: Graphene nanopore

The results from He et al. [6] provide an excellent test of our theory, because the authors simulate a set of charged graphene nanopores which are similar to the selectivity filters of biological channels. The pore diameters are comparable to these filters although the length is smaller \( \sim 0.4 \)nm; meanwhile different charges are added with the intention of reproducing \( K^+ \) or \( Na^+ \) selectivity. These charges are either carbonyl groups with a negative partial oxygen charge or
carboxylate groups with a permanent charge of $-1q$. We shall consider all three pores: the four carboxyl (4CO), four carboxylate (4COO) and three carboxylatey (3COO). Both 4CO and 4COO demonstrate linear responses and so are compared with our linearised expression of current (13); and so the non-Ohmic 3COO pore is compared with our kinetic equation (15). The system was set up with constant temperature 300K and mixed KCl and NaCl solutions of 0.5M each, and Cl$^-$ does not permeate the pore. Thus we can consider the system as having one binding site, and competing single cation permeation because the charging parameter is relatively small $U_c \sim 9$ to 27$kT$ when an averaged value in the pore of $\epsilon_w = 9$ is used. Fitting is provided in Fig. 4 with the fitting parameters given in the caption; and we note that the diffusion coefficient in the pore is smaller than in the bulk as expected [29]. In general the fitting is good with a small residual standard error for each data set, which are given as error bars.

### B. Example 2: $K^+$ biological ion channel (KcsA)

KcsA is a bacterial $K^+$ ion channel strongly favouring $K^+$ over Na$^+$ at ratios of 1000:1 [2]. First crystallised in 1998 [30], it represents a model ion channel because of its shared homology with other $K^+$ channels particularly within the narrow selectivity filter. We approximate this filter as a cylinder$^2$ of length 1.2nm and radius 0.15nm. Inside it contains four binding sites due to attractive dipolar carbonyl oxygen atoms (which exclude anions); with a fifth binding site at on its extra-cellular boundary. In contrast to graphene the charging parameter is large and on the order of the bulk hydration energy, $U_c \sim 160kT$ (with $\epsilon_w \sim 9$ although it is likely to be smaller). This indicates strong interactions between the ions and the pore, resulting in highly coordinated motion with fluctuations between 2 and 3 ions in the filter. In the experimental set-up, symmetrical KCl solutions were used with concentrations ranging from 0.25-1.5M, and current-voltage curves were produced. To model this we reduce the state space and consider a two-level cation filter, with the

\[ \frac{\partial \phi}{\partial t} = kT. \]

The energy barriers for mixing, local and global interactions for each concentration starting at 0.25M to 1.5M were: 3.6$kT$, 3.68$kT$, 3.94$kT$, 4.04$kT$, and 4.33$kT$ respectively.

### VII. Conclusions

In summary, we have presented a multi-scale analysis of ionic conduction and selectivity through subnanometer pores. This includes a multi-species statistical theory describing the state space or occupancy of the pore. Our theory accounts for the properties of the pore such as its charge or geometry, the ionic composition of the surrounding solutions, the global interactions inside the pore, the local interactions specifically at the binding site, and any external driving forces. The important results from this theory have been: the explicit demonstration that noise (fluctuations) in the occupancy of the pore are directly proportional to the rate of conduction through the pore, and the derivation of optimal conducting parameters: the tuned fixed charge $\pi^*_i$ and the optimal chemical potential difference $\Delta \mu^*_i$. In addition we have extended the theory far from equilibrium with a set of master equations. Both theories provide excellent agreement when compared with the experimental and simulated data from the graphene nanopore and the biological KcsA channel.

We have also introduced a molecular dynamics investigation into the role of interactions during the permeation of ions through charged and neutral graphene pores. In both systems the geometry was comparable with a narrow pore. This narrow pore produces geometrical dehydration because water molecules are physically prevented from aligning around the ion as they do in the bulk reservoir (see [10]). Hence the fraction of water molecules lost can be calculated, and we find this to be in good agreement with our simulation data particularly for the uncharged pore. The small differences to the charged pore can be attributed to the lack of water-water and water-pore charge interactions. Consequently the effective continuum of water inside the pore must significantly

\[ \frac{\partial \phi}{\partial t} = kT. \]

2Although we note that structural fluctuations are of order 0.075nm [31].
differ to the bulk reservoir with a smaller effective dielectric coefficient $\varepsilon_d$. We confirm this through comparisons of the PMFS and Coulomb’s law.

To extend the applicability of our theory, we aim to explore the importance of coordinated motion which by describing it as the motion of a quasi-particle [16], [33], [34], with the expectation that it can lead to an improved continuous model of the current through the pore based on a coupled set of differential Chapman-Kolmogorov equations. In addition we would also like to note that the distribution functions (4) are related to the BBGKY hierarchy of kinetic equations [35]; and we plan to further explore this connection. Finally, we believe that the application of our methods will lead to new insights in both technology and biophysics.

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