The Mechanism of Fast-Gate Opening in ClC-0

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ClC-0 is a chloride channel whose gating is sensitive to both voltage and chloride. Based on analysis of gating kinetics using single-channel recordings, a five-state model was proposed to describe the dependence of ClC-0 fast-gate opening on voltage and external chloride (Chen, T.-Y., and C. Miller. 1996. J. Gen. Physiol. 108:237–250). We aimed to use this five-state model as a starting point for understanding the structural changes that occur during gating. Using macroscopic patch recordings, we were able to reproduce the effects of voltage and chloride that were reported by Chen and Miller and to fit our opening rate constant data to the five-state model. Upon further analysis of both our data and those of Chen and Miller, we learned that in contrast to their conclusions, (a) the features in the data are not adequate to rule out a simpler four-state model, and (b) the chloride-binding step is voltage dependent. In order to be able to evaluate the effects of mutants on gating (described in the companion paper, see Engh et al. on p. 351 of this issue), we developed a method for determining the error on gating model parameters, and evaluated the sources of this error. To begin to mesh the kinetic model(s) with the known CLC structures, a model of ClC-0 was generated computationally based on the X-ray crystal structure of the prokaryotic homolog ClC-ec1. Analysis of pore electrostatics in this homology model suggests that at least two of the conclusions derived from the gating kinetics analysis are consistent with the known CLC structures: (1) chloride binding is necessary for channel opening, and (2) chloride binding to any of the three known chloride-binding sites must be voltage dependent.

INTRODUCTION

Members of the CLC family orchestrate the movement of chloride necessary for proper neuronal, muscular, cardiovascular, and epithelial function (Jentsch et al., 2002, 2005; Uchida and Sasaki, 2005; Sile et al., 2006). Defects in these proteins are directly responsible for human diseases of muscle, kidney, bone, and brain (Lloyd et al., 1996; Simon et al., 1997; Cleiren et al., 2001; Kornak et al., 2001; Pusch, 2002; Haug et al., 2003; Naesens et al., 2004). ClC-1, which is highly expressed in mammalian skeletal muscle, is responsible for restoring the resting potential of the cell after muscle contraction (Lehmann-Horn and Jurkat-Rott, 1999; Colding-Jørgensen, 2005). Hence mutations in this channel can lead to myotonia, a condition where voluntary contraction of muscles is only slowly reversible (Lehmann-Horn and Jurkat-Rott, 1999; Colding-Jørgensen, 2005; Jurkat-Rott and Lehmann-Horn, 2005). Many of the myotonia-inducing mutations in ClC-1 cause shifts in voltage-dependent gating (Pusch, 2002). Understanding the voltage dependence of CLC gating is thus of great physiological relevance.

ClC-0, found in Torpedo electric rays, was the first CLC to be discovered and characterized (White and Miller, 1982; Hanke and Miller, 1983; Pusch et al., 1995, 1999; Chen and Miller, 1996; Chen and Chen, 2001; Pusch, 2004; Traverso et al., 2006), and have unprecedented single-channel conductance higher than any other CLC yet studied (∼10 pS), it is the most thoroughly studied CLC and serves as the family prototype (Maduke et al., 1999; Estevez and Jentsch, 2002; Pusch, 2004; Chen, 2005; Jentsch et al., 2005; Dhani and Bear, 2006; Dutzler, 2006). The functional form of ClC-0 is a dimer in which each subunit forms its own pore (Ludewig et al., 1996; Middleton et al., 1996). One gating mechanism, termed “slow” gating, closes both pores simultaneously. A second gating mechanism, termed “fast” gating, governs the independent opening and closing of each pore. Both slow and fast gating are sensitive to transmembrane voltage, chloride, and pH (Miller, 1982; Hanke and Miller, 1983; Pusch et al., 1995, 1999; Chen and Miller, 1996; Chen and Chen, 2001; Pusch, 2004; Traverso et al., 2006), and have unprecedented mechanisms in which the permeant ion plays a key role. Slow gating is energetically coupled to the transmembrane chloride gradient (Richard and Miller, 1990); fast gating voltage dependence arises from the movement of the permeant ion through the transmembrane field (Pusch et al., 1995; Chen and Miller, 1996). Hence gating, permeation, and chloride binding are all tightly coupled in ClC-0.

An interesting feature of the fast gate of ClC-0, revealed by the voltage dependence of its opening rate constant, is that it can be activated by either hyperpolarization or depolarization. These two gating pathways

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can be distinguished further since only the depolarization-activated pathway is sensitive to external chloride concentration.

Analysis of single-channel data led Chen and Miller (1996) to propose a five-state model (Scheme 2) for fast-gate opening that explains the effects of voltage and chloride on the opening rate constant. We aimed to gain insight into what structural changes occur during the steps in this model by examining a series of mutants with altered gating and determining how specific steps in the model are affected (Engh et al., 2007). As a first step toward this goal, we used macroscopic recordings to investigate the effects of voltage and external chloride on the gating kinetics of wild-type ClC-0. While our data display all the same features as those published by Chen and Miller (1996), the use of a different method (global fitting) to fit the data to the theoretical models leads to a substantially different interpretation. By performing a thorough error analysis of both our data and those of Chen and Miller (1996), we show the limitations of fits to the five-state model, and provide an estimate of the uncertainty of these results. We conclude that a simpler four-state model is sufficient to explain the data and that the chloride-binding step is depolarization activated, not voltage independent as previously proposed.

**MATERIALS AND METHODS**

**Channel Expression**

We used a ClC-0 construct in a plasmid derived from the pBlueScript vector (Stratagene) (Jentsch et al., 1990; Maduke et al., 1998), which contained the point mutation C212S. This mutation removes voltage-dependent slow-gate inactivation and has no other measurable effect on ClC-0 function (Lin et al., 1999). Plasmids were linearized with FspI (New England Biolabs), cleaned using the DNA Clean and Concentrator-5 (Zymo Research), and transcribed in vitro using the mMessage mMACHINE T3 RNA-polymerase transcription kit (Ambion). RNA was dissolved in RNAase-free water (Invitrogen) containing 0.77 U/μL SUPERase-In (Ambion).

**Excised Patch Recording**

Defolliculated *Xenopus* oocytes were injected with 27.5 nl RNA at ~1 mg/ml and incubated at 16°C for 2–5 d before recording. Data were then collected from excised inside-out patches using Axopatch 200B and pClamp software. Before patching, the vitelline membrane was removed manually with the oocyte bathed in internal (bath) solution (in mM: 110 NMDG, 110 HCl, 5 MgCl\(_2\), 10 HEPES, 1 EGTA, brought to pH 7.3 using NaOH). Electrical contact between the recording chamber and the ground electrode was made via agarose bridges. Recording electrodes were pulled from 100-μm calibrated pipettes (VWR), polished to 0.2–1.5 MΩ, and filled with external solution. All external (pipette) solutions contained 1 mM EGTA, 10 mM HEPES, and were brought to pH 7.3 using NaOH. For the seven external chloride concentrations used (in mM: 5, 15, 30, 65, 110, 310, 610), the concentration of the other components are shown in Table I. All solutions were sterilized using 0.2-μm filters.

**Measuring Junction Potentials**

Having different solutions on either side of the patch causes significant junction potentials. To accurately correct for these junction potentials we measured them independently. We compared the junction potentials we measured with those calculated using the JPcalc feature of Clampex 8.1 software (Fig. S1, available at http://www.jgp.org/cgi/content/full/jgp.200709759/DC1). When the difference between internal and external solution is extreme (e.g., for external solutions with 5, 300, or 600 mM chloride; internal has 120 mM chloride), there is a significant difference between the measured junction potential and the calculated junction potential. These differences are reduced if activities of the most abundant ions are entered into JPcalc instead of concentrations. We corrected the command voltages using measured junction potentials.

**Determining Open Probability**

We used a voltage protocol designed to obtain the apparent open probability (P\(_o\)) as a function of voltage. (This protocol was also used to determine the opening and closing rate constants, as described in the legend of Fig. 2.) In this protocol, illustrated in Fig. 1, a 50-ms prepulse to a positive voltage (at least +50 mV) is followed by a 170-ms test pulse, starting at least +40 mV above the prepulse voltage and decrementing by 20 mV, and then a 170-ms tail pulse to −100 mV. The prepulse voltage and the range of the test-pulse voltages were adjusted so that the P\(_o\) could be seen to plateau at either end of the voltage range used.

Since the −100-mV tail pulse provides a common driving force to compare the relative number of open channels in each preceding test pulse, we used the current at the beginning of the tail pulse (the instantaneous tail current) to calculate the steady-state open probability reached during each test pulse. The instantaneous tail current, I, was determined by fitting each tail-current decay to a single exponential, and the apparent open probability was calculated as P\(_o\) = I/I\(_{max}\) where I\(_{max}\) is the maximum instantaneous tail current. The gating parameters P\(_{min}\) (the minimum apparent open probability), z (the effective gating charge), and V\(_c\) (the midpoint of the voltage-activation curve) were calculated by fitting the apparent open probability (solid lines) to an equation of the form P\(_o\) = P\(_{min}\) + (1 – P\(_{min}\))/(1 + exp[−2z(V − V\(_c\))/RT]), and are listed in Table II.

**Determining Opening and Closing Rate Constants**

The approach we used to determine opening (α) and closing (β) rate constants is described in the legend of Fig. 2. This method becomes inaccurate under the following circumstances: (a) when the current decay is so fast that it is difficult to separate from the capacitive transient, making it difficult to accurately determine 1/τ (i.e., at very negative voltages and/or low external chloride); (b) when P\(_o\) is small and therefore difficult to accurately determine (i.e., at very negative voltages); (c) when the amplitude of the current decay is small so it is difficult to determine 1/τ accurately (i.e., at higher voltages); (d) when there is little current present.

**TABLE I**

*External (Pipette) Solutions*  

<table>
<thead>
<tr>
<th>Component</th>
<th>5</th>
<th>15</th>
<th>30</th>
<th>65</th>
<th>110</th>
<th>310</th>
<th>610</th>
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<tbody>
<tr>
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<td>100</td>
<td>100</td>
<td>100</td>
<td>100</td>
<td>100</td>
<td>300</td>
<td>600</td>
</tr>
<tr>
<td>Glutamate</td>
<td>95</td>
<td>85</td>
<td>70</td>
<td>35</td>
<td>0</td>
<td>0</td>
<td>0</td>
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<td>0</td>
<td>0</td>
<td>0</td>
</tr>
</tbody>
</table>

Each column represents a separate pipette solution. The top row lists the concentration of each of the components of the solution.
passing through the patch, such that the signal-to-leak ratio is small and neither \(1/\tau\) nor \(P_c\) can be accurately determined.

To avoid inaccuracies arising from the circumstances mentioned above, we investigated any patches that gave \(\alpha\) values that appeared to be outliers. We excluded data from patches where the raw data had any obvious problems, such as (a) the prepulse current continually drifted down between sweeps, (b) the test-pulse current at high voltages drifted down radically over the course of the pulse, and (c) the tail currents from all the sweeps did not approach the same final value. These three measures resulted in the exclusion of <6% of the 256 records taken (for each patch we took one to four records, with each record testing a different set of voltages). We also excluded data from individual sweeps where (a) there were large and biased residuals for the single exponential fit to the test-pulse or tail-pulse decay, (b) noise in the record was so large that the instantaneous tail current was less than zero, resulting in a \(P_c\) less than zero, or (c) the resulting \(\alpha\) value failed a Dixon test for outliers. We used the Dixon tests and 95% confidence cutoff values as described in Outliers in Statistical Data, tests N7–N13 (Barnett and Lewis, 1994). These three measures resulted in the exclusion of <2% of the 2185 sweeps recorded. Finally, after averaging the \(\alpha\) values for each condition (voltage, external chloride concentration), we excluded conditions where there were extremely large error values (mean/SEM > 0.4), since the data taken at these conditions are likely to be too variable to be trusted. This last measure resulted in the exclusion of <3% of the 246 conditions tested.

The use of macroscopic recordings to derive \(\alpha\) values is based on two important assumptions. First, this method assumes that the maximum \(P_c\) is equal to 1, which has been shown using single-channel recordings over a wide range of external chloride concentrations (4–600 mM; Chen and Miller, 1996). Second, this method assumes that there are only two conductance levels (open and closed) for each pore in the channel. Single-channel recordings have shown that this is the case for wild type (Miller, 1982; Hanke and Miller, 1983; Richard and Miller, 1990; Ludewig et al., 1996; Middleton et al., 1996; Accardi and Pusch, 2003) and K149L (Zhang et al., 2006). These assumptions seem reasonable because the wild-type \(\alpha\) values we measured using macroscopic patch recordings show the same dependence on voltage and external chloride as those obtained from single-channel recordings (as reported by Chen and Miller, 1996).

**Global Fitting**

We used multivariate fitting procedures to fit the opening rate constant as a function of voltage and external chloride activity to equations for either the four-state or the five-state model. To do this fits, we used the Levenberg-Marquardt algorithm as implemented in Igor (Wavemetrics). To improve fitting in Igor, we fit \(\log(\alpha)\) rather than \(\alpha\), weighted the fits by 1/SEM, and wrote user-defined functions such that the parameters sought were all of the same order of magnitude. We subsequently transformed the parameters to their actual order of magnitude.

**Determining the Activity of Chloride in our Solutions**

The analyses and calculations shown were done using chloride activities; similar results were obtained using chloride concentrations. To calculate the chloride activity, we had to consider the effects of other ions present in the solution. The major ions in our solutions were NMDG, chloride, and in the cases of low chloride concentration, glutamate. Unfortunately, the activity coefficient for NMDG-CI has not been measured. Since the difference between the activity coefficient of potassium chloride and choline chloride is small over the relevant range of concentrations (Fig. S2), we assume that the difference between the activity coefficient of NMDG-CI and choline chloride is small. Since NMDG is closer in size to choline than to potassium, we used the activity coefficient of choline chloride to calculate the chloride activity for each of our solutions (activity = activity coefficient * concentration).

**Sequence Alignment and Structural Modeling**

To construct a three-dimensional model of the atomic structure of wild-type CIC-0, we followed the homology modeling procedure described below, based on the structure of the prokaryotic proton/chloride exchanger ClC-ec1 (Dutzler et al., 2003). The underlying assumption of this approach is that the true structure of ClC-0 (whether wild type or mutant) does not depart substantially from that of its homolog, especially in regions of high sequence conservation. We consider this a reasonable assumption based on results showing the reliability of the ClC-ec1 structure as a guide to CIC-0 (Estevez et al., 2003; Lin and Chen, 2003; Engh and Maduke, 2005).

First, to identify the optimal sequence alignment between CIC-0 and ClC-ec1 (Fig. S3), we obtained a consensus prediction from (a) a multiple-sequence alignment including all known members of the CLC family, performed with Chustal-W (Thompson et al., 1994), and (b) a profile-to-profile alignment, performed with HMAP (Petrey et al., 2003). The latter method includes both primary sequence information, derived from a multiple-sequence alignment of the output of a PSI-BLAST search, as well as secondary-structure information, derived from the known structure of the ClC-ec1 and from a consensus prediction for CIC-0 (using NET, PSIPRED, and PHD).

Based on the resulting alignment and the structure of the ClC-ec1 dimer (PDB ID 1OTS), an initial model of wild-type CIC-0 was built using the molecular modeling software CHARMM (Brooks et al., 1983), preserving the three-dimensional structure of the backbone of the prokaryotic homolog, and excluding insertions of more than five amino acids. The conformation of conserved or sterically similar side chains (e.g., Glu and Gln) was also preserved, in addition to the chloride ions bound to ClC-ec1 in the and S sites; otherwise, the prediction algorithm SCWRL (Camatessu et al., 2003) was employed to determine the most plausible side chain conformations. Subsequently, a stepwise scheme was employed to energy minimize the atomic model, where structural constraints based on the degree of conservation were gradually released. These energy minimizations also were performed with CHARMM, using the CHARMM22 all-atom force field (MacKerell et al., 1998), a 14-Å cutoff for the nonbonded interactions, and a distance-dependent dielectric constant.

For comparison, we also built a model inspired by the structure of the E148Q mutant of ClC-ec1 (Dutzler et al., 2003) using an analogous procedure to that described above. In this structure an additional chloride ion is bound to the S sites; in the wild-type structure, this site is occupied by the side chain of E148. In our ClC-0 homology model inspired by the E148Q structure, the S site may be occupied by chloride, while the side chain of E166 (the E148 equivalent) is protonated and rotated around the \(\chi_i\) dihedral angle. After energy minimization of this model, E166 forms a hydrogen bond with C212.

**Electrostatic Binding Energies and Transmembrane Potential**

The electrostatic contribution to the free energy of binding of each chloride ion \(i\) to the remainder of the protein–ion complex \(p\) was computed according to the expression:

\[
\Delta G_i = \Delta G_{p}^\phi - (\Delta G_{p} + \Delta G_i^\phi),
\]

where \(\Delta G_{p}^\phi\), \(\Delta G_{p}\), and \(\Delta G_i^\phi\) represent the electrostatic energies of the full complex, the protein and chloride ions other than \(i\), and the isolated ion \(i\) in solution, respectively. For each of these systems, the electrostatic energy is defined as:

\[
\Delta G = \frac{1}{2} \sum_{j}^N \Phi(r_j),
\]
where $g_i$ and $r_i$ denote the charge and location of each of the $N$ atoms in the protein–ion complex and $\Phi(r)$ is the electrostatic potential evaluated at those positions. To compute $\Delta G^{\text{m}}(r)$, $\Delta G^{\text{m}}(r)$, and $\Delta G(r)$, we employed the Poisson-Boltzmann equation in its weak-field, linear approximation:

$$V \cdot [\varepsilon(r)V\Phi(r)] - \kappa^2(r)\Phi(r) = -\frac{4\pi}{\varepsilon_0} \sum_{j=1}^{N_i} q_j \delta(r-r_j), \quad (3)$$

where the position-dependent quantities $\varepsilon(r)$ and $\kappa(r)$ are the dielectric constant and the Debye-Hückel ionic screening factor, respectively. The dielectric distribution $\varepsilon(r)$ was partitioned into different regions: the region delimited by the solvent-accessible surface of the protein–ion complex (constructed with a 1.4 Å probe radius) was assigned a dielectric constant $\varepsilon_{\text{water}}$, equal to 4.8, or 20 in independent calculations; the membrane region, represented by a continuous slab of overlapping, uncharged pseudo-atoms in which the protein–ion complex is embedded, was assigned a dielectric constant $\varepsilon_{\text{membrane}} = 2$; and the remainder of the system was assigned a water-like dielectric constant $\varepsilon_{\text{solvent}} = 80$. Similarly, the screening factor $\kappa(r)$ due to the ionic solution (in this case of monovalent ions, e.g., NaCl, with concentration $C = 150$ mM) was partitioned as

$$\kappa(r) = \begin{cases} \frac{8\pi\varepsilon_0 N_A C}{10^5 - k_B T} & \text{if } r \in \text{bulk}, \\ 0 & \text{if } r \notin \text{bulk}. \end{cases} \quad (4)$$

where $N_A$ and $k_B$ are Avogadro’s and Boltzmann’s constants, respectively, and $T$ the temperature; here, the “bulk” region comprises the portion of the system that is accessible to the ions in solution, i.e., the region outside the solvent-accessible surface of protein and membrane, minus a 2 Å-wide Stern layer.

Following Roux (1997) and Nonner et al. (2004), we also used the Poisson-Boltzmann framework to analyze the spatial dependence of the electrostatic contribution to chloride binding ($\Delta G_{\text{e}}$) resulting from an externally applied transmembrane voltage $V_{\text{m}}$, via the expression:

$$V \cdot [\varepsilon(r)V\Phi_{\text{m}}(r)] - \kappa^2(r)\Phi_{\text{m}}(r) - \theta(r)V = 0, \quad (5)$$

where $\theta(r)$ is a step function equal to one on one side of the membrane, and equal to zero on the other side.

To solve the Poisson-Boltzmann equation, we used the grid-based, finite-difference solver PBEQ (Im et al., 1998) implemented in version C32A2 of the CHARMM molecular modeling software (Brooks et al., 1983), with a modification that allows for the use of pseudo-atoms with an arbitrary dielectric constant. Atomic charges were obtained from the CHARMM22 force field (MacKerell et al., 1998), alongside the optimized atomic radii derived by Nina et al. (1997).

Grid focusing and translational averaging were implemented in all calculations of the electrostatic binding energy in order to minimize discretization errors without significantly increasing the computational cost. The dimensions of the initial grid were $116 \times 74 \times 80$ Å, with a grid-point spacing of 1.0 Å. Boundary conditions were interpolated from this calculation and used for a second grid, of dimensions $104 \times 62 \times 78$ Å with a grid-point spacing equal to 0.5 Å. The focusing procedure was repeated for eight different locations of the protein–membrane system relative to the center of the three-dimensional grids (shifted 0–0.25 Å in each direction); the $\Delta G_{\text{e}}$ values reported hereafter are averages over these independent calculations, as well as over the two protein subunits. For the calculation of $\Phi_{\text{m}}(r)$, the electrostatic potential was computed along a putative permeation pathway across the channel, for a transmembrane voltage $V_{\text{m}} = 200$ mV and a grid-point spacing of 0.5 Å. For each position along this pathway, the values of $V/V_{\text{m}}$ reported are averages of $\Phi_{\text{m}}$ over nine sites shifted on the plane of the membrane by $\pm 1$ Å, as well as over both protein subunits. Standard deviations of the averages of $\Delta G_{\text{e}}$ and $V/V_{\text{m}}$ are shown as error bars in Fig. 6.

Online Supplemental Material

The supplemental material is available online at http://www.jgp.org/cgi/content/full/jgp.200709759/DC1. Fig. S1 shows the activity coefficients for chloride in the patching solutions used. Fig. S3 shows the alignment of ClC-0 and ClC-ec1 used to create the ClC-0 homology model. The supplemental material also includes a large part of the kinetic analysis, including a detailed description of Method 1 (Fig. S4), a comparison of the results obtained using Method 1 and Method 2 (Fig. S5), and a section on determining the interdependency of parameters in the gating model (Fig. S6). Also included are the results of calculations analyzing the effects of chloride occupancy on chloride binding affinity, in which the electrostatic contribution to chloride binding ($\Delta G_{\text{e}}$) to various sites in the pore is determined under various occupancy conditions (Fig. S7).

RESULTS

The Effect of Voltage and Chloride on Fast-Gate Open Probability

To assess the effects of voltage ($V$) and external chloride concentration ([Cl]$_{\text{ext}}$) on fast gating, we measured currents through macroscopic excised inside-out patches from *Xenopus* oocytes expressing ClC-0. To do this we used the pulse protocol described in Materials and methods and illustrated in Fig. 1A (top left). Examples of the current response to this protocol at different external chloride concentrations (Fig. 1A) illustrate that over this wide range of chloride concentration, tail currents can be used to derive the apparent open probability (Fig. 1B). As has been previously reported (Pusch et al., 1995; Chen and Miller, 1996), we found that increasing external chloride concentration activates the fast gate, causing a leftward shift in the plot of the open probability as a function of voltage.

The Effect of Voltage and Chloride on Fast-Gate Kinetics

The results from our macroscopic recordings of wild-type ClC-0 show similar features to those obtained by Chen and Miller using single-channel recordings: (a) the opening rate constant ($\alpha$) as a function of voltage gives a nearly V-shaped curve similar to a chevron plot (Chen and Dill, 1998), where for the right-hand arm of the channel (voltages less negative than $\sim -175$ mV, depending on external chloride concentration), opening is dominated by a depolarization-activated process; and for the left-hand arm of the channel (voltages more negative than $\sim -175$ mV), opening is dominated by a hyperpolarization-activated process (Fig. 2, A and B); (b) only the right-hand arm of the curve is shifted by external chloride (Fig. 2, A and B); (c) the closing rate constant shows a log-linear dependence on voltage.
Figure 1. Effects of external chloride on gating. (A) Voltage pulse protocol (top left) and current responses in 5, 110, and 310 mM external chloride. (B) Tail-current analysis was used to determine the apparent open probability ($P_o$) as a function of voltage ($V$) (see Materials and methods) ($n = 5–26$ for external chloride concentrations <600 mM, and $n = 2–5$ for 610 mM). These plots were fit as described in Materials and methods (solid lines), and gating parameters were derived (Table II).

and thus appears to be governed by only one process (Fig. 2 C). Fast-gate closing is activated by hyperpolarization and is slightly sensitive to changes in external chloride concentration.

Methods of Fitting the Data
Chen and Miller (1996) proposed both a four-state model (Scheme 1) and five-state model (Scheme 2) to describe the dependence of fast-gate opening on voltage and chloride. These two models are similar in that they both include a chloride-bound (C-Cl) and a chloride-free (C) closed state, and an open state (O).

In both models the channel can open from either closed state. The difference between these models is not the states that the channel can visit but the connectivity between states. In the five-state model the channel can open after the chloride-binding step via a hyperpolarization-activated or a depolarization-activated pathway. In contrast, in the four-state model the channel must open through a depolarization-activated pathway once the chloride-binding event has occurred. Mathematically, however, these two models are very similar (see online supplemental material, available at http://www.jgp.org/cgi/content/full/jgp.200709759/DC1) and both have six free parameters. We fit our data to both the four- and five-state gating models and in the process of doing so, discovered that the method of fitting the data greatly affects the results and the estimation of errors. To better compare our results to those of Chen and Miller, we reanalyzed their data as well as ours using several methods, as described below.

Chen and Miller’s Method of Fitting (Method 1)
As described in detail in the supplemental material, Chen and Miller used a two-step method (Method 1) to

<table>
<thead>
<tr>
<th>[Cl]$_{ext}$</th>
<th>$V_o$</th>
<th>$z$</th>
<th>$P_{min}$</th>
<th>$n$</th>
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<tr>
<td>5</td>
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<td>$0.070 \pm 0.011$</td>
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<td>$0.90 \pm 0.01$</td>
<td>$0.070 \pm 0.008$</td>
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</tr>
<tr>
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<td>$0.051 \pm 0.020$</td>
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</table>
fit the opening rate constant to the four- and five-state models, as a function of voltage and chloride. We used Method 1 to fit our data and Chen and Miller’s to both the four- and five-state models. To best compare these two datasets we needed to compare the parameter values resulting from the fits to the kinetic models. We found that the range of voltages included in the fit significantly affected the resulting parameter values (Fig. 3). This suggests that the error on the parameter values is large and that the datasets cannot be compared without accurate error estimates. Unfortunately, since Method 1 is a two-step process, it is difficult to estimate the error for the parameter values determined using this method.

Fitting the Data Using Global Fits (Method 2)
Because of the difficulty in estimating errors when using Method 1, we used a different method of fitting the data that allowed us to estimate the error in the parameter values calculated. This enabled us to determine whether our dataset yields significantly different results from that of Chen and Miller. In this approach, Method 2, the data were globally fit, allowing all six parameters to be free. When Method 2 is used, both Chen and Miller’s dataset (solid lines in Fig. 2 B and Fig. S5 C) and our dataset (solid lines in Fig. 2 A and Fig. S5 F) fit well to both the four- and five-state models.

Four-State versus Five-State model
Since Chen and Miller (1996) rejected the four-state model because it did not fit their data well, we were surprised that when using Method 2, both our data and those of Chen and Miller fit well to the four-state model. The reason for this discrepancy is that when Method 1 is used (as in Chen and Miller, 1996), the four-state model parameters derived predict that within the range of the data (a) there will be an intersection point for all the curves of α versus V (where each curve represents a different chloride concentration); and (b) once the chloride binding is saturated, α will vary log-linearly with voltage (instead of having a chevron shape). However, when Method 2 is used to analyze the data, the resulting parameters are such that the intersection point occurs outside the range of the data, and α is no longer predicted to vary log-linearly with voltage at saturating chloride concentrations. Therefore at this point there is no strong evidence to suggest whether the four-state or the five-state model is more accurate.
Although both models have six free parameters, and therefore mathematically are of equal complexity, the four-state model is conceptually simpler. Since it is important to compare our results for wild-type ClC-0 to those published by Chen and Miller, we will continue to include analysis using the five-state model since that was the model Chen and Miller chose to use for their final analysis. In the Discussion, however, we will only refer to the four-state model and its parameters.

Error Analysis on Global Fits
Since Method 2 is a one-step fitting process where all the data from a particular dataset are fit at once, we were able to readily estimate the errors in parameter values calculated by this global fitting method. To estimate these errors, we used a procedure to test how changing a given parameter value affects the goodness of fit to the five-state model. To estimate how much each parameter value could vary and still yield a good fit to the data, we used the following procedure: (a) the parameter of interest was held at a test value and $\alpha(V, [\text{Cl}]_{\text{ext}})$ was globally fit with the other five parameters to be free, shown as solid lines in Fig S5, C and F. Method 1 was performed including either all the voltages (all) or some subset of the voltages (sub). For Method 1, the error bars show the standard deviation calculated by the fitting algorithm (for fits shown in Fig S4, B and D); for Method 2, the error bars show the 95% confidence limits determined as described in the text. For comparison, the first column shows the parameter values ($z_1^*, z_2^*$, and $z_c$) published by Chen and Miller (pub. val.); values for $\alpha_1^*(0)$, $\alpha_2^*(0)$, and $K_c(0)$ were not reported.

$$SS = \sum (\alpha_m - \alpha_l)^2,$$

(6)
where for every \( \alpha(V, [Cl]_{ext}) \) point, \( \alpha_m \) is the measured \( \alpha \) value, and \( \alpha_f \) is the \( \alpha \) value calculated by the fit; (c) steps (a) and (b) were repeated over a large range of test values. For each of the six parameters in the model, steps (a)–(c) were performed for a wide range of test values. Plotting the sum of squares (SS) as a function of each parameter (Fig. 4) results in minima near the best-fit value for the fit where all parameters were free (compare values in Fig. 5, column 5, to parameter values at the SS minima in Fig. 4). The plots in Fig. 4 show how markedly the goodness of fit depends on each parameter, allow us to be sure that we sampled a large enough range of parameter values, and show that there are no other minima. We also used the plots in Fig. 4 to estimate confidence intervals for each parameter that was determined by globally fitting \( \alpha(V, [Cl]_{ext}) \).

To obtain 95% confidence bars, we determined the sum of squares at 95% confidence:

\[
SS(95\%) = SS_{BF}(1 + ((F * P) / (N - P))),
\]

where \( SS_{BF} \) is the sum of squares for the best fit when all parameters are free, \( P \) is the number of free parameters in the model, \( N \) is the number of data points being fit, and \( F \) is the critical value of the F distribution for a p-value of 0.05 (Motulsky and Christopoulos, 2004). For each parameter determined by global fitting, 95% confidence intervals were obtained as the two parameter values that result in a sum of squares equal to the sum of squares at 95% confidence (see the intersection of the two lines in plots shown in Fig. 4). These 95% confidence intervals are shown as error bars in Figs. 3 and 5, and are much larger than the standard error values calculated by the Igor fitting algorithm.

With these error bars it is easy to see that when Method 2 is used, there is no significant difference between the parameter values that result for our dataset and that of Chen and Miller, for both the four-state (Fig. 3, compare the last two columns in each chart) or five-state models (Fig. 5, compare the last two columns in each chart). Thus, despite the differences in data collection techniques, internal chloride concentrations used, and voltage range analyzed, the parameters for ClC-0 opening are not significantly different.

The Value of \( z_c \)

The error bars in Figs. 3 and 5 illustrate that some of the parameter values calculated using Method 2 are significantly different from those previously published (for each graph compare the first column to the last two). The difference that is conceptually important is that in our analysis of both our data and those of Chen and Miller, using Method 2 and either the four- or five-state models (Fig. 3 and 5, respectively), we found that \( z_c \) is significantly different from zero, between \(-0.9\) and \(-0.1\). This demonstrates that chloride binding during fast-gate opening (the \( K_c \) step) is voltage dependent, contrary to the conclusions of Chen and Miller (1996). Since \( z_c \) represents the voltage dependence of \( K_c \), the equilibrium constant for chloride dissociation, the \( z \) value for chloride association, \( z_a \), is greater than zero (\( z_a = -z_c \)). This suggests that the \( K_c \) step involves the binding of chloride from the extracellular milieu to site(s) within the transmembrane electric field.

Since this is an important result, we wanted to evaluate the probability that the chloride-binding step is not voltage dependent. To evaluate this possibility, we tested to see how holding \( z_c \) at a value near to zero would affect the other results. We used Method 2 and globally fit both our data and those of Chen and Miller to the five-state
Engh et al.

While holding \( z_c \) constant at 0.08, the value published by Chen and Miller (1996), this approach resulted in the fits shown as dashed lines in Fig. 2 (A and B), and the parameter values in Fig. 5 (column 2 and column 3 in each graph). Since holding \( z_c \) to 0.08 gives reasonable-looking fits, but significantly affects some of the parameter values, we needed to know if this alternate method of fitting could be more valid.

To this end, we used an F-test to compare the two methods for fitting the data to the five-state model: the method where only five parameters are free (\( z_c \) is held) and the method where all six parameters are free. The F-test takes into account the difference in goodness of fit and the difference in degrees of freedom. In this test, described by Motulsky and Christopolous (2004), an F ratio is calculated:

\[
F = \frac{(SS_{null} - SS_{alt})/DF_{null} - DF_{alt})}{(SS_{null} / DF_{null})},
\]

where \( SS \) is the sum of squares for the best fit to the data, and \( DF \) is the degrees of freedom ((number of data points) − (number of free parameters)). In this case, the fit using six free parameters is the more complex “alt” model and the fit using five free parameters is the less complex “null” model. We calculated F ratios of 43 and 121 for Chen and Miller’s dataset and ours, respectively. These correspond to p-values of \( 2 \times 10^{-9} \) and \( 3 \times 10^{-23} \). The p-values indicate the likelihood that the simpler model, with only five free parameters, is the better model to use. This statistic indicates that the method of using a global fit with all six parameters free was a better method to use. Therefore it is unlikely that \( z_c \) is zero, and from these data we conclude that chloride binding during fast gating is voltage dependent.

The Interdependence of Parameters in the Five-State Model

Since holding \( z_c \) at 0.08 changed the values calculated for the other parameters in the global fit, we realized that the parameters are likely to be interdependent. This could explain why some of the error bars on the gating parameters are so large (last two columns in each graph of Figs. 3 and 5). Interdependence between parameters means that there are not enough features in the relationship between the opening rate constant as a function of voltage and external chloride to allow all six parameters in the gating model(s) to be uniquely determined. Therefore, there are a number of fits that are all equally good.

As described in the supplemental material, we used the data from the error analysis to determine whether any of the parameters in the four- or five-state models are interdependent. This resulted in the fits shown as dashed lines in Fig. 2 (A and B), and the parameter values in Fig. 5 (column 2 and column 3 in each graph). Since holding \( z_c \) to 0.08 gives reasonable-looking fits, but significantly affects some of the parameter values, we needed to know if this alternate method of fitting could be more valid.

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values are at least partially due to these interdependencies. The results of this analysis are similar when the four-state model is used.

Chloride Binding during the $K_c$ Step

We wondered whether our finding that $K_c$ is voltage dependent, in combination with information from the X-ray crystal structures of ClC-ec1, could tell us about what chloride movements occur during the $K_c$ step. To address this question, we used information from the X-ray crystal structures of ClC-ec1 to try to learn more about the electrostatics of chloride binding in the pore. First, we used the structures of ClC-ec1 (Dutzler et al., 2003) in conjunction with sequence alignments (see Materials and methods and Fig. S3) to construct two homology models for ClC-0. One homology model was based on the structure of wild-type ClC-ec1, which has been proposed to resemble the ClC-0 closed state; the other homology model was based on the structure of the E148Q ClC-ec1 mutant, which has been proposed to resemble a ClC-0 conducting state. The difference between these two X-ray crystal structures is that in the wild-type structure the $S_{\text{ext}}$ chloride-binding site is occupied by the side chain of E148, and in the E148Q mutant structure the 148 side chain is rotated out and a chloride is bound in $S_{\text{ext}}$. Both of these homology models were made with minimal adjustments to the ClC-ec1 structures (see Materials and methods).

Thus, the overall shape of the pore was preserved, as were the three chloride-binding sites detected in that structure, namely $S_{\text{int}}$, $S_{\text{cen}}$, and $S_{\text{ext}}$ (moving from the cytoplasm to the extracellular space). We then performed an analysis of the electrostatics of chloride binding to the homology model, using the Poisson-Boltzmann framework.

Our analysis of the homology models suggests that chloride binding to all three sites is voltage dependent. This is illustrated in Fig. 6 A, which shows the fraction of transmembrane potential ($V/V_{\text{m}}$) a chloride ion has to traverse to reach each of the three binding sites along a putative permeation pathway. (Results from the two homology models are barely distinguishable.) The $V/V_{\text{m}}$ plot shows that a chloride ion coming from the extracellular milieu would have to traverse a significant portion of the transmembrane potential to reach $S_{\text{ext}}$, $S_{\text{cen}}$, or $S_{\text{int}}$. This means that chloride binding to any of these sites, relative to the bulk, would be voltage dependent. Since $K_c$ is voltage dependent, having a $z_2$ between $-0.1$ and $-0.9$, it is likely to involve chloride binding to one of these three sites.

**TABLE III**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>$z_1$</th>
<th>$z_2$</th>
<th>$z_3$</th>
<th>$K_c(0)$</th>
<th>$K_c$</th>
</tr>
</thead>
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<tr>
<td>$\alpha_1$</td>
<td>+</td>
<td>+</td>
<td>-</td>
<td>0.1</td>
<td>-</td>
</tr>
<tr>
<td>$\alpha_2$</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>0.1</td>
<td>-</td>
</tr>
<tr>
<td>$\alpha_3$</td>
<td>+</td>
<td>+</td>
<td>+</td>
<td>0.1</td>
<td>+</td>
</tr>
</tbody>
</table>

Contour plots such as those shown in Fig. S6 were used to judge the interdependence of pairs of parameters when $\alpha(V, [Cl]_{\text{ext}})$ is globally fit (using Method 2) to the four- and five-state models. + indicates interdependence, ++ indicates strong interdependence, – indicates no interdependence.

**Electrostatic Contribution to the Free Energy of Chloride Binding**

Previous experimental work has shown that both conductance and gating show anomalous mole fraction effects in CLC channels (Pusch et al., 1995; Rychkov et al., 1998). Since this suggests that chloride–chloride repulsion in the pore plays a significant role in ClC-0 function, we wanted to characterize the energetic coupling between chloride ions in the pore and evaluate how this coupling could affect fast gating. Using the ClC-0 homology model inspired by the E148Q ClC-ec1 structure, we calculated the electrostatic contribution to the free energy of chloride binding ($\Delta G_c$) to $S_{\text{int}}$, $S_{\text{cen}}$, $S_{\text{ext}}$ (Eq. 1). Fig. 6 B shows that for chloride binding to any of the three sites, the chloride binding affinity is less when there are two chlorides already in the pore (squares) than when there are no chlorides already bound (circles). The results of this electrostatic analysis are therefore consistent with the anomalous mole fraction data and previous in silico analyses (Cohen and Schulten, 2004; Gervasio et al., 2006). As described in the supplemental material, in terms of electrostatics, the effect of having the E148 side chain (E166 in ClC-0) rotated into the $S_{\text{cen}}$ site appears to be comparable to that of having a chloride bound in this site. Hence the calculated $\Delta G_c$ values for $S_{\text{int}}$ and $S_{\text{cen}}$ are similar for the homology model based on the wild-type ClC-ec1 structure and the homology model inspired by the E148Q mutant when it has a chloride bound in $S_{\text{ext}}$. As discussed below, comparison of $\Delta G_c$ for different sites and under different conditions helps to evaluate models for chloride movement during fast gating.

**DISCUSSION**

Here we have shown that using macroscopic patch recordings we can reproduce the effects of voltage and external chloride on the open probability and gating kinetics reported by Chen and Miller (1996) based on
the transmembrane voltage along a hypothetical permeation pathway, with the transmembrane voltage of $-200 \text{ mV}$. Shown here is the fraction of the transmembrane voltage along a hypothetical permeation pathway, with $S_{\text{int}}$, $S_{\text{cen}}$, $S_{\text{ext}}$ at sites 10, 14, and 16, respectively. (B) The CIC-0 homology model was used to calculate the electrostatic contribution to the free energy of chloride binding ($\Delta G_b$) to each site in the pore (starting from the cytoplasmic side: $S_{\text{int}}$, $S_{\text{cen}}$, $S_{\text{ext}}$) with either zero (circles) or two (squares) chlorides already bound in the pore. The dielectric constant used for the protein was 4. In both A and B, lines connect data points for ease in viewing, and error bars represent an estimate of the discretization error arising from the grid-based Poisson-Boltzmann calculations (see Materials and methods).

Figure 6. Chloride binding and movement in the pore. (A) The CIC-0 homology models created as described in Materials and methods were used to calculate the electrostatic potential at various positions in the pore ($V/V_{\text{mem}}$) under an externally applied transmembrane voltage of $-200 \text{ mV}$. Shown here is the fraction of the transmembrane voltage along a hypothetical permeation pathway, with $S_{\text{int}}$, $S_{\text{cen}}$, $S_{\text{ext}}$ at sites 10, 14, and 16, respectively. (B) The CIC-0 homology model was used to calculate the electrostatic contribution to the free energy of chloride binding ($\Delta G_b$) to each site in the pore (starting from the cytoplasmic side: $S_{\text{int}}$, $S_{\text{cen}}$, $S_{\text{ext}}$) with either zero (circles) or two (squares) chlorides already bound in the pore. The dielectric constant used for the protein was 4. In both A and B, lines connect data points for ease in viewing, and error bars represent an estimate of the discretization error arising from the grid-based Poisson-Boltzmann calculations (see Materials and methods).

their single-channel recordings. We have also shown that the method of analysis greatly affects the gating parameter values derived from these data. We propose an alternate methodology (Method 2) that involves globally fitting the data with all parameters free. This approach allows for (a) a quantitative determination of the error in the calculated parameter values, which is essential for comparing different datasets, and (b) an assessment of the interdependence of the parameters in the gating model. The interdependence between the parameters probably makes a large contribution to the error in the parameter values determined. Our analysis using Method 2 shows that (a) our data and those of Chen and Miller yield similar gating parameter values; (b) the four- and five-state models fit the data equally well; and (c) the chloride-binding step that leads to channel opening is voltage dependent. This last finding changes our conceptualization of the gating process, as described in the next section.

A Model for Chloride Movement during Gating

Our ultimate goal is to understand the structural changes that occur during fast gating in CIC-0. As a first step toward a structural model for gating, we aimed to use previously published results and our new findings to develop a model for how chloride moves during depolarization-activated fast-gate opening: $\text{C} \rightarrow \text{C} \cdot \text{Cl} \rightarrow \text{O}$. Given that there are at least three chloride-binding sites in CIC-0, there are many possibilities for how chloride could move during depolarization-activated fast-gate opening.

$K_c$ Is Likely to Involve Binding to $S_{\text{cen}}$, $S_{\text{int}}$, or $S_{\text{ext}}$

Our homology model analysis suggests that the chloride-binding sites $S_{\text{cent}}$, $S_{\text{int}}$, and $S_{\text{ext}}$ in CIC-0 are all significantly into the transmembrane field (Fig. 6 A). (It is important to note that the transmembrane voltage profile is primarily sensitive to the extent to which the pore can be permeated by high-dielectric solvent. Therefore, for any reasonable model based on the CIC-ec1 structures, it is improbable that any of the binding sites mentioned above would be equipotential with the bulk environment.) However, it is conceivable that the extracellular vestibule is wider in CIC-0 than in CIC-ec1. We wondered whether the extracellular vestibule of CIC-0 could be so wide that $S_{\text{ext}}$ is not within the transmembrane field, such that chloride from the extracellular milieu binding to this site would not be voltage dependent. To examine the impact that widening the extracellular vestibule may have on the transmembrane profile, we built a CIC-0 homology model lacking helix L, a helix that lines the extracellular vestibule. In spite of the much greater solvent accessibility of this region of the protein that results from the deletion of this helix, the transmembrane voltage at the chloride-binding sites was found to be essentially unchanged relative to our original model (unpublished data). Since $K_c$ is voltage dependent, it is therefore reasonable to conclude based on this analysis that the $K_c$ step is likely to involve binding to $S_{\text{cen}}$, $S_{\text{int}}$, or $S_{\text{ext}}$.

There remains the question of whether $K_c$ could involve chloride binding to some yet undiscovered site extracellular to $S_{\text{ext}}$. Since $z_c$ is between $-0.1$ and $-0.9$, such a site would have to be at least $10\%$ into the transmembrane electric field, which corresponds to a position $\sim 6$–$7$ Å into the pore. One possibility is the $S_{\text{ext}}$ site proposed by Faraldo-Gómez and Roux (2004), which is in the vicinity of H401, or alternatively, the sites proposed by Yin et al. (2004) and Bostick and Berkowitz (2004), near K165. These proposals are worth noting in view of electrophysiological data showing that K165 mutations affect both permeation and fast gating (Lin and...
Chen, 2000), and that H401 mutations affect permeation (Zhang et al., 2006). However, mutation at H401 does not affect gating, and the effects of mutations at K165 and H401 could occur through allostery. All these predictions of a fourth chloride-binding site are based on computational analyses of the ClC-ec1 structures, and no structural data is as yet available that confirms the existence of a such a site, including the more recent crystallographic studies (Lobet and Dutzler, 2006; Accardi et al., 2006). Thus it is unclear whether H401 and K165 line an additional extracellular chloride-binding site or whether the mutations exert their effects via some other mechanism. Therefore in our models for chloride movement during gating, we have only included the three chloride-binding sites observed in the X-ray crystal structures.

### Chloride Movement during the $\alpha_2^*$ Step

Analysis of ClC-0 gating kinetics indicates that the step following chloride binding, the $\alpha_2^*$ step (referring now only to the four-state model), is voltage dependent. Given the previously observed anomalous mole fraction effects on $V_o$ (the voltage at the midpoint of the voltage-activation curve) (Pusch et al., 1995), this voltage dependence has been thought to be due to the movement of chloride through the pore. The calculated $V/V_m$ profile (Fig. 6 A) suggests that a chloride ion has to traverse a significant portion of the transmembrane field to travel between binding sites in the pore, and therefore all steps in gating that require chloride movements between sites are going to be voltage dependent. This leads us to conclude that the $\alpha_2^*$ step could involve chloride movement between any of the sites in the pore.

### Hypothetical Models for Chloride Movement

We used the evidence discussed so far to generate models for chloride movement during depolarization-activated gating. After considering a wide range of possibilities, we arrived at a set of more likely models (Fig. 7 A) based on the following criteria: (a) only $S_{\text{ext}}$, $S_{\text{cen}}$, and $S_{\text{int}}$ are considered; (b) the $K$ step involves chloride from the extracellular milieu binding to one of these sites; (c) the $\alpha_2^*$ step can involve depolarization-activated chloride movement (solid lines) or binding (dashed lines) (but does not have to); it could involve depolarization-activated conformational changes that do not affect chloride occupancy; (d) $S_{\text{ext}}$ becomes occupied during the course of gating, and must be occupied for the channel to remain open. The lines connecting states indicate likely transitions, given the simplifying assumption that only one chloride can bind per step. It is important to note that the open state (O) depicted does not represent all the possible chloride occupancy states explored once chloride has started to conduct through the pore, but rather the minimum chloride occupancy required for the channel to remain open.

### The Two Most Likely Models

Analyses of ClC-ec1 structures suggest that the likely models shown in Fig. 7 A can be narrowed down further. These analyses come in two forms. First, Lobet and Dutzler (2006) obtained and analyzed ClC-ec1 crystal structures under a variety of anion concentrations in order to estimate the apparent binding affinity of chloride for each of the three sites observed. Second, we estimated the electrostatic contribution to the chloride binding affinity for each of the three sites for all the possible occupancy states, (Fig. 6 B; Fig. S7). Both these analyses suggest the following: (a) ClC-0 has a sufficiently high affinity for chloride such that the state where no chloride is bound is unlikely under our experimental conditions.
Engh et al. 347

(5–610 mM extracellular chloride, 120 mM intracellular chloride). If the chloride-free state (the top C state in Fig. 7 A) is eliminated from the list of likely possibilities, then by extension we eliminate the C·Cl and O states connected only to this C state. (b) Chloride appears to have a higher affinity for $S_{\text{cen}}$ than $S_{\text{int}}$, such that the second C state (from the top) in Fig. 7 A is more likely than the third, and the $\alpha_2^*$-step transition that involves movement of chloride from $S_{\text{cen}}$ to $S_{\text{int}}$ is unlikely.

Therefore, based on the available information, we can winnow down the possible models in Fig. 7 A to the two most likely options. These two options are shown in Fig. 7 B: Model 1, (top) C has one chloride bound in $S_{\text{cen}}$, C·Cl has two chlorides bound, and O has three chlorides bound; Model 2, (bottom) C has two chlorides bound in $S_{\text{cen}}$ and $S_{\text{int}}$, C·Cl has three chlorides bound, and O also has three chlorides bound. In Model 1, we have modified the first option so that each step is simplified and only involves either the binding of chloride to $S_{\text{cen}}$ or the downward movement of chloride between adjacent binding sites in the pore. The C·Cl state consists of three microstates in equilibrium (outlined by a dashed line).

These two models are fundamentally different from one another. In Model 1, each of the steps can be simply the chloride binding or chloride movement shown. Although additional protein structural rearrangements are allowed, this model does not require any additional conformational changes other than that the E166 side chain rotate out of $S_{\text{ext}}$ to allow the binding of chloride to this site. The only requirement for opening is that all sites are occupied with chloride. In contrast, both steps in Model 2 require protein structural movements: the $K_c$ step involves the rotation of E166 and the binding of chloride to $S_{\text{ext}}$; the $\alpha_2^*$ step has to involve some structural change that does not change chloride occupancy but makes conduction through the channel easier. For example, this transition could involve the lowering of barriers to chloride movement between the chloride-binding sites. In Model 2 there are two requirements for opening: (1) that all three sites are occupied with chloride and (2) that some additional protein conformational change has occurred so that the channel can conduct. Additional protein conformational changes in CLC gating have been proposed by others (Accardi and

**Figure 7.** Possible models for chloride movement during depolarization-activated fast gating. Evidence from previously published papers and our kinetic and homology model analysis was used to narrow down the possible models for chloride movement during depolarization-activated fast-gate opening to those shown (see text). The channel begins in the closed state C, binds chloride to reach the closed state C·Cl, and then goes through another depolarization-activated transition to reach the open state O. This is depicted at the top of both A and B. $K_c$ is the apparent equilibrium constant for chloride dissociation, and $\alpha_2^*$ is the rate constant for the second step, C·Cl → O. In A, depicted below each of these states (C, C·Cl, and O) are their possible chloride occupancy states, with empty circles depicting vacant chloride binding sites, and filled circles depicting filled chloride binding sites. Lines between these chloride-occupancy states indicate transitions corresponding to either C → C·Cl or C·Cl → O. Dashed lines indicate C·Cl → O transitions that involve binding of chloride to the pore. Evidence from our analysis of gating kinetics and the ClC-0 homology models further narrowed the possibilities to those shown in B. In Model 1, the dashed line outlines the microstates that are all part of C·Cl. $K_1$, $K_2$, and $K_3$ are microscopic equilibrium constants, and $k_4$ is a microscopic rate constant. In Model 2, the second step has to involve some conformational change that opens the channel but does not change chloride occupancy.
the cytoplasmic side, it is likely, if these apparent affinities are in the millimolar range, and we used 120 mM chloride based on the apparent chloride affinities calculated by Lobet and Dutzler (2006). Since for all three sites the apparent affinities Lobet and Dutzler calculate are in the millimolar range, and we used 120 mM chloride on the cytoplasmic side, it is likely, if these apparent affinities hold for ClC-0, that both $\Delta_{\text{int}}$ and $\Delta_{\text{cen}}$ were filled before the channel opening that we measured.

**Conclusion**

In conclusion, we have shown that the method of analyzing the dependence of the opening rate constant on voltage and chloride affects the conclusions drawn about the mechanism of fast gating in ClC-0. We have shown that both the four- and five-state gating models fit the data, and that the accuracy of the calculated gating parameter values is compromised by the interdependence of parameters. Despite the uncertainty in determining gating parameter values, we can conclude that chloride binding during gating is a voltage-dependent process, and must involve a binding site at least 10% into the transmembrane field. In this regard, our homology model analysis shows that any of the three known chloride-binding sites could play a role in gating. By considering results from our kinetic analysis, previously published results, and available structural information, we propose two possible models for chloride movement during depolarization-activated fast gating. Although more information is needed to determine which of these two models is more likely, understanding the assumptions in each of these models provides fodder for future experiments. Both the methods of analysis and error estimation described here and the structural models proposed provide a framework for evaluating how mutations in ClC-0 affect fast gating.

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