Efficacy of Synaptic Inhibition Depends on Multiple, Dynamically Interacting Mechanisms Implicated in Chloride Homeostasis

Nicolas Doyon¹², Steven A. Prescott³, Annie Castonguay¹², Antoine G. Godin¹, Helmut Kröger⁴, Yves De Koninck¹²⁺

¹Division of Cellular and Molecular Neuroscience, Centre de recherche Université Laval Robert-Giffard, Québec, Québec, Canada, ²Department of Psychiatry & Neuroscience, Université Laval, Québec, Québec, Canada, ³Department of Neurobiology and Pittsburgh Center for Pain Research, University of Pittsburgh, Pittsburgh, Pennsylvania, United States of America, ⁴Department of Physics, Université Laval, Québec, Québec, Canada

Abstract

Chloride homeostasis is a critical determinant of the strength and robustness of inhibition mediated by GABA_{A} receptors (GABA_{AR}s). The impact of changes in steady state Cl⁻ gradient is relatively straightforward to understand, but how dynamic interplay between Cl⁻ influx, diffusion, extrusion and interaction with other ion species affects synaptic signaling remains uncertain. Here we used electrodiffusion modeling to investigate the nonlinear interactions between these processes. Results demonstrate that diffusion is crucial for redistributing intracellular Cl⁻ load on a fast time scale, whereas Cl⁻ extrusion controls steady state levels. Interaction between diffusion and extrusion can result in a somato-dendritic Cl⁻ gradient even when KCC2 is distributed uniformly across the cell. Reducing KCC2 activity led to decreased efficacy of GABA_{AR}-mediated inhibition, but increasing GABA_{AR} input failed to fully compensate for this form of disinhibition because of activity-dependent accumulation of Cl⁻. Furthermore, if spiking persisted despite the presence of GABA_{AR} input, Cl⁻ accumulation became accelerated because of the large Cl⁻ driving force that occurs during spikes. The resulting positive feedback loop caused catastrophic failure of inhibition. Simulations also revealed other feedback loops, such as competition between Cl⁻ and pH regulation. Several model predictions were tested and confirmed by [Cl⁻] imaging experiments. Our study has thus uncovered how Cl⁻ regulation depends on a multiplicity of dynamically interacting mechanisms. Furthermore, the model revealed that enhancing KCC2 activity beyond normal levels did not negatively impact firing frequency or cause overt extracellular K⁺ accumulation, demonstrating that enhancing KCC2 activity is a valid strategy for therapeutic intervention.

Introduction

In the central nervous system, fast inhibition is mediated by GABA_{A} and glycine receptor-gated Cl⁻ channels (GABA_{AR} and GlyR). Influx of Cl⁻ through these channels produces outward currents that cause hyperpolarization or prevent depolarization caused by concurrent excitatory input (i.e., shunting) [1,2]. Hyperpolarization and shunting both typically reduce neuronal spiking. However, Cl⁻ influx through GABA_{AR} necessarily increases [Cl⁻], which in turn causes depolarizing shifts in the Cl⁻ reversal potential (E_{Cl}) [3,4]. As the Cl⁻ gradient is depleted and E_{Cl} rises, the efficacy of GABA_{AR}-mediated control of spiking is compromised [5]. Therefore, mechanisms that restore the transmembrane Cl⁻ gradient are crucial for maintaining the efficacy of GABA_{AR}-mediated inhibition.

Cation-chloride cotransporters (CCCs) play a key role in maintaining the Cl⁻ gradient across the membrane [6,7]. Most relevant to neurons are the Na⁺-K⁺-2Cl⁻ cotransporter (NKCC1), which normally mediates uptake of Cl⁻ [8], and the K⁺-Cl⁻ cotransporter, isoform 2, (KCC2), which normally extrudes Cl⁻. Interestingly, a reduction in KCC2 expression and/or function is involved in the pathogenesis of several neurological disorders, including epilepsy and neuropathic pain [9–15]. Motivated by the clinical relevance of hyperexcitability caused by changes in KCC2 activity, conductance-based compartmental models have been used to study how changes in E_{Cl} influence inhibitory control of neuronal spiking [5].

E_{Cl} can change as a result of altered KCC2 expression or activity [7,16,17]. E_{Cl} can also change dynamically, on a fast time scale, as a result of Cl⁻ flux through GABA_{A} receptors, particularly in small structures like distal dendrites [2,4,18]. If E_{Cl} changed only slowly, it could be reasonably approximated as static relative to other neuronal processes occurring on a faster time scale; however, since E_{Cl} changes rapidly, it may interact in

Citation: Doyon N, Prescott SA, Castonguay A, Godin AG, Kröger H, et al. (2011) Efficacy of Synaptic Inhibition Depends on Multiple, Dynamically Interacting Mechanisms Implicated in Chloride Homeostasis. PLoS Comput Biol 7(9): e1002149. doi:10.1371/journal.pcbi.1002149

Editor: Abigail Morrison, University of Freiburg, Germany

Received December 24, 2010; Accepted June 11, 2011; Published September 8, 2011

Copyright: © 2011 Doyon et al. This is an open-access article distributed under the terms of the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.

Funding: This research was supported by the Natural Sciences and Engineering Research Council (NSERC) of Canada (YDeK) and by the National Institutes of Health, RO1 NS063010 (SAP). ND was supported by postdoctoral fellowships from NSERC and the Neurophysics training program funded by the Canadian Institutes of Health Research. SAP is a Rita Allen Foundation Scholar in Pain and the 53rd Mallinckrodt Scholar. AC was supported by a postdoctoral fellowship from Fonds de la Recherche en Santé du Québec (FRSQ). Y De K is a Chercheur National of the FRSQ. The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Competing Interests: The authors have declared that no competing interests exist.

* E-mail: yves.dekoninck@crulrg.ulaval.ca
Author Summary

Fast synaptic inhibition relies on chloride current to hyperpolarize the neuron or to prevent depolarization caused by concurrent excitatory input. Both scenarios necessarily involve chloride flux into the cell and, thus, a change in intracellular chloride concentration. The vast majority of models neglect changes in ion concentration despite experimental evidence that such changes occur and are not inconsequential. The importance of considering chloride homeostasis mechanisms is heightened by evidence that several neurological diseases are associated with deficient chloride extrusion capacity. Steady state chloride levels are altered in those disease states. Fast chloride dynamics are also likely affected, but those changes have yet to be explored. To this end, we built an electrodiffusion model that tracks changes in the concentration of chloride plus multiple other ion species. Simulations in this model revealed a multitude of complex, nonlinear interactions that have important consequences for the efficacy of synaptic inhibition. Several predictions from the model were tested and confirmed with chloride imaging experiments.

potentially complex ways with important neuronal processes like synaptic integration. To investigate those interactions, one must treat [Cl\textsuperscript{−}] as a dynamical quantity evolving in space and time. The spatio-temporal dynamics of [Cl\textsuperscript{−}] depend on several factors, including GABA\textsubscript{A}R-mediated Cl\textsuperscript{−} flux, longitudinal diffusion within dendrites and the soma, and CCC activity. Furthermore, Cl\textsuperscript{−} dynamics involve complex non-linear interactions with other ion species, which have been overlooked by previous models [19]. To understand how these dynamical processes interact with each other, we built an electrodiffusion model that monitors intra- and extracellular concentrations of several ion species (Cl\textsuperscript{−}, Na\textsuperscript{+}, K\textsuperscript{+}, Ca\textsuperscript{2+}, HCO\textsubscript{3}−, H\textsuperscript{+}, HPO\textsubscript{4}\textsuperscript{2−}, H\textsubscript{2}PO\textsubscript{4}−) across neuronal compartments (see Fig. 1A–C). Our model revealed several consequences of impaired Cl\textsuperscript{−} extrusion on neuronal function, including a positive feedback loop between intracellular Cl\textsuperscript{−} accumulation and excitatory activity or spiking that can lead to catastrophic failure of inhibition. Several predictions of the model were confirmed by direct measurement of [Cl\textsuperscript{−}]\textsubscript{i} by fluorescence lifetime imaging microscopy (FLIM).

Results

\(E_{\text{Cl}}\) and \(E_{\text{GABA}}\) depend non-linearly on KCC2 activity and synaptic input

Past experiments have established that Cl\textsuperscript{−} extrusion via KCC2 plays a crucial role in maintaining the values of \(E_{\text{Cl}}\) and \(E_{\text{GABA}}\) below the resting membrane potential [20], but they have not established how KCC2 activity relates quantitatively to \(E_{\text{Cl}}\) and \(E_{\text{GABA}}\) in particular under conditions of ongoing, distributed synaptic input. Therefore, as a first step, we varied KCC2 activity and measured the impact on \(E_{\text{Cl}}\) and \(E_{\text{GABA}}\) (measured at the soma) in a model neuron receiving a fixed level of background excitatory and inhibitory synaptic input (Fig. 1D). Values of \(E_{\text{Cl}}\) and \(E_{\text{GABA}}\) in middle and distal dendrites are described by similar curves shifted to slightly more depolarized values (data not shown) consistent with the somato-dendritic gradient described below. This is important since neurons in vivo are bombarded by synaptic activity [21], but it remains unclear how this may affect \(E_{\text{Cl}}\) and, in turn, be affected by \(E_{\text{Cl}}\). Consistent with qualitative experimental findings [9,22,23], both reversal potentials underwent depolarizing shifts as KCC2 activity was reduced, with \(E_{\text{Cl}}\) approaching the mean membrane potential (Fig. 1D). Notably, \(E_{\text{GABA}}\) was less negative than \(E_{\text{Cl}}\), especially at high values of KCC2 activity, consistent with \(E_{\text{GABA}}\) depending jointly on [Cl\textsuperscript{−}], and [HCO\textsubscript{3}−], [24]. However, unlike the large depolarizing shift in \(E_{\text{Cl}}\) caused by reducing KCC2 activity, increasing KCC2 activity beyond its normal value caused only a marginal hyperpolarizing shift in \(E_{\text{Cl}}\), which approached the K\textsuperscript{+} reversal potential (\(E_{\text{K}}\)) near -90 mV.

This is consistent with KCC2 normally operating near equilibrium. Hence, while activation in KCC2 activity can cause strong reduction of inhibition, excess KCC2 activity has a limited influence on the strength of inhibition, insofar as we assume that strength of GABA\textsubscript{A}R-mediated inhibition is a function of the value of \(E_{\text{GABA}}\).

Thus, in addition to validating our model, this first set of simulations revealed an interesting nonlinear relationship between KCC2 activity and \(E_{\text{Cl}}\). However, we expected that \(E_{\text{Cl}}\) should depend not only on KCC2, but also on factors like GABA\textsubscript{A}R input – this was the main motivation for developing an electrodiffusion model. As a preliminary test, we varied the rate of inhibitory synaptic input together with KCC2 activity. Results show that \(E_{\text{Cl}}\) underwent a depolarizing shift, the magnitude of which depended on KCC2 activity, as the rate of inhibitory input increased (Fig. 1E). At a normal KCC2 level, increasing the activation rate of GABA\textsubscript{A}R synapses from 0.2 to 5 Hz drove \(E_{\text{Cl}}\) up by only 7 mV, whereas the same change in activation rate drove \(E_{\text{Cl}}\) up by 24 mV when KCC2 activity was decreased to 33% of its normal value. Thus, KCC2 activity not only controls baseline \(E_{\text{Cl}}\) but it also determines how stably \(E_{\text{Cl}}\) is maintained when the Cl\textsuperscript{−} load is increased by synaptic input. Tonic inhibition due to activation of extrasynaptic GABA\textsubscript{A} receptors by ambient GABA can also contribute to intracellular Cl\textsuperscript{−} accumulation and depolarize \(E_{\text{Cl}}\).

To test the impact of tonic inhibition, we performed simulations with and without this form of inhibition. Results obtained with and without tonic inhibition were qualitatively the same (Fig. 1E).

To test experimentally the impact of the level of KCC2 activity on intracellular Cl\textsuperscript{−} accumulation, we loaded neurons in primary cultures (>21 days in vitro; DIV) with MQAE and measured changes in [Cl\textsuperscript{−}]\textsubscript{i} using FLIM. FLIM measurements have the advantage of being unbiased by the amount of indicator from cell to cell (Fig. 2A, B). To test the variability between measurements as well as shielding the measurements from changes in cell volumes [25]. We first bath applied the GABA\textsubscript{A}R agonist muscimol to trigger Cl\textsuperscript{−} influx through GABA\textsubscript{A}R channels. We then applied various concentrations of furosemide or VU 0240551 for 20 minutes to block KCC2 activity. In the presence of Cl\textsuperscript{−} load through activated GABA\textsubscript{A} channels, application of furosemide or VU 0240551 led to dose-dependent Cl\textsuperscript{−} accumulation (Fig. 2C, D), in agreement with the predictions of simulations (cf. Fig. 1D). At high doses, furosemide can antagonize both KCC2 and NKCC1; however, at > 21 DIV, hippocampal neurons in culture are generally thought to fully express KCC2 but to no longer express NKCC1 [6]. To test this, we used bumetanide at a concentration (50 μM) where it selectively blocks NKCC1. Administration of bumetanide to cells exposed to muscimol caused Cl\textsuperscript{−} import. To test for this, we blocked KCC2 with the recently developed selective blocker VU 0240551 [25]. Further addition of bumetanide after KCC2 blockade had no effect on [Cl\textsuperscript{−}]\textsubscript{i}, confirming absence of significant NKCC1-mediated transport in these neurons (Fig. 2D). These results indicate a) significant KCC2 co-transport in > 21 DIV hippocampal neurons in culture, maintaining [Cl\textsuperscript{−}]\textsubscript{i} at a low level, and b) that both furosemide and...
VU 0240551 could be used under these conditions to selectively block KCC2-mediated transport.

With the importance of nonlinear interaction between GABAAR activity and KCC2 activity for intracellular Cl$\textsuperscript{-}$ regulation thus established, we moved onto more detailed analysis of how Cl$\textsuperscript{-}$ flux impacts the efficacy of synaptic inhibition. Transmembrane Cl$\textsuperscript{-}$ gradient may vary between cellular compartments depending on the spatial distribution of synaptic input and cotransporter activity.

Spatial variation in $E_{\text{Cl}}$ (or $E_{\text{GABA}}$) between cellular compartments has been observed in several experiments [20,26–29] but it is not typically accounted for in conventional neuron models. While a longitudinal, axo-somatodendritic [Cl$\textsuperscript{-}$] gradient could be due to differentially distributed cotransporter activity, it could also arise from intense focal GABAAR-mediated input. To test the latter scenario, we simulated high frequency GABAAR-mediated input to a single dendritic synapse and measured [Cl$\textsuperscript{-}$] at different distances from the synapse at different times after the onset of input (Fig. 3A). Under the conditions tested, a GABAAR synapse activated at 50 Hz produced a longitudinal [Cl$\textsuperscript{-}$] gradient of 50 μM/μm, which extended as far as 60 μm and could yield changes in $E_{\text{GABA}}$ on the order of 5 mV within 200 ms (Fig. 3A). There were only subtle differences between centripetal and centrifugal diffusion (i.e., toward or away from soma, respectively; Fig. 3B). According to these data, if a GABAAR synapse receives sustained high frequency input, [Cl$\textsuperscript{-}$] will increase near that synapse, influencing $E_{\text{GABA}}$ at the original synapse as well as at more distant synapses.

---

Figure 1. Summary and initial validation of model. A. Schematic of model neuron showing geometry and compartment dimensions. B. Summary of ion flux mechanisms included in the model (see Methods for details). Diffusion in the extracellular space is not depicted. C. Sample traces of membrane potential together with [K$\textsuperscript{+}$]$_o$ measured in the extracellular shell surrounding the soma and [Cl$\textsuperscript{-}$] measured in the soma (black) and in a dendrite (red). This is only a subset of ion species whose concentrations were continuously monitored in all compartments, and from which reversal potentials were continuously updated. D. As predicted, reducing KCC2 below its "normal" level (100%) caused large depolarizing shifts in $E_{\text{Cl}}$ and $E_{\text{GABA}}$, whereas increasing KCC2 up to 400% above normal caused only minor hyperpolarizing shifts. Simulation includes background synaptic input with $f_{\text{inh}}$ = 0.8 Hz and $f_{\text{exc}}$ = 0.2 Hz/synapse. The dashed line represents the mean value of membrane potential averaged over 200 s. E. Reversal potentials also depended on the rate of GABAAR input, which dictates the Cl$\textsuperscript{-}$ load experienced by the neuron. Increasing $f_{\text{inh}}$ caused a depolarizing shift in $E_{\text{Cl}}$, the extent of which increased when KCC2 was decreased. For these simulations, $f_{\text{inh}}/f_{\text{exc}}$ ratio was fixed at 4 and $f_{\text{inh}}$ was varied from 0.05 Hz to 4.8 Hz. Dashed lines represent results from simulations performed with tonic GABA conductances while solid lines represent simulations performed without it.

doi:10.1371/journal.pcbi.1002149.g001
nearby synapses. This was further investigated by placing a “test” GABA<sub>A</sub> synapse (activated at 5 Hz) at varying distances from the original GABA<sub>A</sub> synapse (activated at 50 Hz). Both synapses were activated simultaneously. As predicted, E<sub>GABA</sub> at the test synapse was affected by other GABA<sub>A</sub>-mediated input on the same dendrite as far away as 50 μm (Fig. 3C top), or even farther when KCC2 activity was reduced. However, interactions also depended on synapse position relative to the neuron topology; for instance, synapses in relatively close proximity but located on different primary dendrites exhibited little if any interaction (Fig. 3C bottom), consistent with the soma acting as a sink that clamps [Cl<sup>-</sup>]<sub>i</sub>.

Under in vivo conditions, neurons are known to be constantly bombarded by synaptic input [30]. We therefore tested whether this synaptic noise affects [Cl<sup>-</sup>]<sub>i</sub> differently depending on the cellular compartment. We performed simulations in the presence or absence of KCC2 activity and in the presence or absence of synaptic noise. Simulations of distributed ongoing synaptic input with KCC2 distributed uniformly across the cell compartments yielded a clear somato-dendritic [Cl<sup>-</sup>]<sub>i</sub> gradient (Fig. 4A black). In contrast, in the absence of simulated synaptic noise, there was no significant somato-dendritic [Cl<sup>-</sup>]<sub>i</sub> gradient despite the presence of KCC2 (Fig. 4A green). Lack of a significant somato-dendritic [Cl<sup>-</sup>]<sub>i</sub> gradient was also observed in the reverse scenario, i.e. in the presence of synaptic noise but without KCC2 (Fig. 4A red). Thus, a significant somato-dendritic [Cl<sup>-</sup>]<sub>i</sub> gradient can exist when there is ongoing Cl<sup>-</sup> influx, redistribution of that Cl<sup>-</sup> load via diffusion, and Cl<sup>-</sup> extrusion by KCC2. This clearly demonstrates that differential extrusion, i.e. inhomogeneous KCC2 density (see below), is not necessary for inhomogeneous transmembrane Cl<sup>-</sup> gradients.

To test the predictions made by the model, we used FLIM to measure [Cl<sup>-</sup>]<sub>i</sub> in MQAE-loaded neurons in culture (Fig. 4B). To mimic distributed Cl<sup>-</sup> influx across the dendritic tree, we exposed the cultures to 100 μM muscimol. FLIM measurements indicated significant changes in [Cl<sup>-</sup>]<sub>i</sub> in MQAE-loaded hippocampal neurons (26 DIV). The mean intensity of MQAE fluorescence within the cell bodies 1 & 2 was significantly different (left), which could be interpreted as indicating different levels of [Cl<sup>-</sup>]<sub>i</sub> or different dye uptake and accumulation between the two cells. The lifetime maps of the same cells are shown in the micrograph on the right. Note how, in contrast to intensities, the fluorescence lifetime of both cells was not significantly different indicating that there were no difference in [Cl<sup>-</sup>]<sub>i</sub> between the two cells. Values are mean ± S.D. of all pixels in each cell body. B. Measurements of MQAE lifetime at different [Cl<sup>-</sup>]<sub>i</sub>, inside the cell body after membrane permeabilization and equilibration with [Cl<sup>-</sup>]<sub>o</sub> at 8, 15 or 20 mM (N = 73 cells/12 coverslips). According to the Stern-Volmer equation: \( t_0 = \frac{1}{t} = 1 + K_{sv}[Cl^-] \). The measured Ksv from these data was 32 M<sup>-1</sup>, consistent with previous reports [86]. C. Effect of increasing concentration of furosemide (to block KCC2) on [Cl<sup>-</sup>]<sub>i</sub>, in cultured neurons exposed to 100 μM muscimol (to evoke a constant Cl<sup>-</sup> load by opening GABA<sub>A</sub>R; N = 75 cells/10 coverslips). D. The selective KCC2 antagonist VU 0240551 caused a dose-dependent significant increase in [Cl<sup>-</sup>]<sub>i</sub> (p<0.05), but bumetamide had no significant (n.s.) effect alone or after blocking KCC2 with VU 0240551, indicating lack of significant NKCC1 transport in these cells (N indicated in each bar = cells/coverslips; ***, p < 0.001).

doi:10.1371/journal.pcbi.1002149.g002
a significant $[\text{Cl}^-]$ gradient along dendrites (Fig. 4B bottom) which was either reduced by bicuculline (Fig. 4B middle and 4C) or blocked by the addition of furosemide or the recently developed more specific KCC2 inhibitor VU 0240551 [25] (Fig. 4B bottom and 4C), consistent with predictions from simulations (cf. Fig. 4A). The small remaining gradient in the presence of furosemide may...
indicate the presence of another chloride transport mechanism not accounted for in the model. Our simulations were based on the assumption of even distribution of KCC2 along the dendrites and this configuration appears to be sufficient to explain the somato-dendritic gradient observed. However, this does not rule out the possibility of a gradient of KCC2 along the dendrites. To test for the presence or absence of such gradient, we sought to perform quantitative fluorescence immunocytochemical analysis of the distribution of KCC2 along dendrites. Measuring KCC2 immunolabeling may not be sufficient, however, to obtain an estimate of the distribution of functional KCC2 because it has recently been suggested that the oligomeric form of KCC2 is the functional one [31,32]. To specifically measure the density of KCC2 dimers along the dendrites we took advantage of a technique we recently developed, entitled Spatial Intensity Distribution Analysis (SpIDA) which allows quantitative measurement of the density and oligomerization of proteins from conventional laser scanning confocal microscopy analysis of immunocytochemical labeling [33,34]. We thus applied SpIDA to analysis of of KCC2 immunostaining of dendrites of the neurons used in the pharmacological experiments described above (Fig. 4). The monomeric quantal brightness was estimated using immunolabeling of KCC2 in neurons that have been in culture for only 5 days, because, at that stage of development, KCC2 has been shown to be essentially monomeric [31]. The monomeric quantal brightness was estimated to be 3.9±0.2 Miu, and was constant along the dendrite of 5-DIV neurons (52 regions from 11 neurons). Using automated intensity binary masks [35], the dendrites of the mature neurons (> 21 DIV) were carefully detected and intensity histograms were generated for each analyzed region and a two-population (monomers and dimers) mixture model was assumed. For each analyzed region, SpIDA was performed on the image of the z-stack (0.5 μm between images) that had the brightest mean intensity in the chosen region. To estimate the true membrane density of KCC2, the final value for each region was averaged over the two adjacent images of the z-stack. A neuron with example regions and their corresponding histogram and SpIDA fit values are presented in Figure 5A,B. The results indicate that the membrane density of KCC2 is constant along the dendrites, at least as far as 200 μm from the center of the cell body (Fig. 5C).

While our experimental results indicate homogeneous distribution along the dendrite length, this does not necessarily apply to all conditions and, in particular, our analysis did not focus on local inhomogeneities, e.g. microdomains. We therefore also sought to determine if longitudinal intracellular Cl⁻ gradients could also arise from inhomogeneous CCC activity at small length scales. For instance, non-uniform distribution of KCC2 at the subcompartent-level might produce local gradients comparable to those

Figure 4. A standing somato-dendritic Cl⁻ gradient is caused by the joint action of KCC2 activity and GABAAR mediated synaptic input. A. Distribution of [Cl⁻] in a modeled dendrite as a function of distance from the soma in the presence and absence of Cl⁻ load due to distributed synaptic activity and of Cl⁻ extrusion through uniformly distributed KCC2. B. Photomicrographs of an example cell loaded with MQAE with lifetime color coding (blue: low Cl⁻ concentration, red: high Cl⁻ concentration). Intracellular Cl⁻ concentration was measured in the presence of muscimol (Musc; 100 μM) and/or bicuculline (Bic; 100 μM) and/or furosemide (Furo; 200 μM) and/or VU 0240551 (VU, 15 μM). Arrows indicate the location where measurements were performed. C. Effect of tonic activation of GABAARs by muscimol on [Cl⁻] in real dendrites as a function of distance from the soma (each data point represent mean ± SEM taken from 10–12 neurons; values from several dendrites were averaged for each cell). Bicuculline and/or furosemide and/or VU 0240551 were added to block Cl⁻ loading and extrusion, respectively. doi:10.1371/journal.pcbi.1002149.g004
observed with synaptic inputs (see Fig. 3). Indeed, clustering of KCC2 has been observed near some synapses [36], but KCC2 near excitatory synapses has been shown to serve a role in scaffolding rather than as a co-transporter [37]. Nevertheless, to test whether subcellular distribution of KCC2 can yield local gradients, we simulated high frequency synapses at 20 μm intervals, between each firing synapses Cl− extrusion through KCC2 was localized at a single point that was placed at different distances from the synapses (Fig. 6A). In all cases, the location of KCC2 had an impact on $E_C$, of $<2$ mV. Thus, our simulations showed that subcompartemental distribution of KCC2 (i.e., inhomogeneities on the spatial scale of 0-10 μm) has little impact on the perisynaptic value of $E_C$.

The results above do not rule out the possibility of inhomogeneities in CCC expression underlying gradients in other cells types, as well as inhomogeneities in the axon initial segment and soma with respect to dendrites. For instance, absence of KCC2 in the axon initial segment (AIS) [9,38], selective expression of the inward Cl− transporter NKCC1 in the AIS [28], or the combination of both expression patterns would be expected to cause $E_C$ to be less negative in the AIS. To test $E_C$ in the AIS and how it impacts neighboring compartment, we simulated different levels of NKCC1 in the AIS in combination with different levels of KCC2 in the soma and dendrites with or without background synaptic input (Fig. 6B–C). NKCC1 expression in the AIS can produce an axo-somatic [Cl−]i gradient, but this gradient does not extend far, if at all, into the dendrites (Fig. 6B). As expected, combining NKCC1 expression in the AIS with synaptic noise (like in Fig. 4A) resulted in a “double gradient” (Fig. 6C right panel).

Thus, simulations in our electrodiffusion model demonstrated that subcellular distribution of GABAAR input and CCC activity can produce spatial inhomogeneities in $E_C$, which should translate into inhibitory input having differing efficacy depending on the location of the synapse. This is true even if KCC2 activity is uniformly distributed in the presence of background GABAAR input. Moreover, focal Cl− influx through one synapse (or a cluster of synapses) can affect the efficacy of neighbouring synapses, although this depends on subcellular localization of those interacting synapses, e.g. proximity to the soma. In contrast, subcompartemental inhomogeneity in KCC2 activity is not sufficient to cause local [Cl−], gradients.

### Diffusion and KCC2 activity determine how robustly the transmembrane chloride gradient is maintained during high frequency synaptic input

Figures 4 and 6 emphasized how spatial variations in [Cl−], can arise from ongoing GABAAR input. To extend these results to include temporal changes in $E_C$, we considered how [Cl−], evolves during stimulus transients. This was motivated by experimental observations that $E_{GABA}$ can rapidly collapse during bursts of GABAAR synaptic events [20,28,39,40]. Activity-dependent changes in $E_{GABA}$ depend on the location of the input: somatic input has less impact on $E_{GABA}$ than dendritic input [4,28]. Simulations in our electrodiffusion model replicated those experimental data (Fig. 7A) as well as results from simpler models [19]. A train of synaptic inputs to the soma produced a small depolarizing shift in $E_{GABA}$, which translated into a small reduction in GABAAR-mediated current. The depolarizing shift in $E_{GABA}$ was greater and occurred increasingly faster for input to progressively more distal dendrites. This was despite the presence of KCC2 (red). Removing KCC2 (black) increased the amplitude and speed of the collapse in Cl− gradient during high frequency input to distal dendrites, but had virtually no impact for input to the soma. The finding that amplitude of the initial synaptic event in each of the compartments was unaffected by removing KCC2 appears to contradict the observation that the standing [Cl−], gradient depends on KCC2 activity (see Fig. 4). We hypothesized that this was due to the absence of ongoing Cl− load caused by the lack of background synaptic activity. We therefore repeated simulations shown in Figure 7A but with background synaptic input (Fig. 7B). As predicted, the initial IPSC amplitude was affected by the KCC2 activity level when background synaptic input was present (compare Fig. 7B and A). These results suggest...
that the rate of local intracellular Cl\(^{-}\) accumulation depends principally on diffusion (which redistributes the intracellular Cl\(^{-}\) load), whereas the extent of accumulation depends on KCC2 activity (which reduces intracellular Cl\(^{-}\) load via extrusion).

To investigate these processes more thoroughly, we systematically varied the intraburst frequency, location of the “test” synapse and KCC2 activity, and we measured the mean IPSC amplitude at the “test” synapse throughout the burst. During high

Figure 6. Inhomogenous CCC distribution can create large-scale, but not fine-scale, intracellular [Cl\(^{-}\)]\(_i\) gradients. A. left: To investigate whether the perisynaptic distribution of KCC2 can produce fine-scale intracellular [Cl\(^{-}\)]\(_i\) gradients, we varied the subcompartmental distribution of KCC2 by concentrating it in a single location in each compartment at varying distances from a bursting synapse. We divided the compartment into 20 1-μm-long sections. Total amount of KCC2 per compartment was constant at 100%. Inhibitory synapses were located at 20 μm from each other and were activated at high frequency. Right: Results show that the subcompartmental distribution has little impact on the perisynaptic value of $E_{Cl}$, which contrasts with the impact of high frequency synaptic input (see Fig. 3) but is consistent with diffusion being responsible for rapid redistribution of intracellular Cl\(^{-}\) load. B. In the absence of synaptic activity, we inserted different levels of NKCC1 activity in the axon initial segment (AIS) and monitored the axo-somato-dentritic [Cl\(^{-}\)]\(_i\) gradient for high (100%) and low (33%) levels of KCC2 activity (uniformly distributed, except in the AIS). Soma corresponds to 0 on x-axis; positive distance extends towards dendrites and negative distance extends towards axon, as summarized on left panel. C. In the presence of background synaptic activity ($f_{inh} = 0.4$ Hz; $f_{exc} = 0.1$ Hz) we simulated different levels of KCC2 activity (uniformly distributed, except in the AIS) and monitored the axo-somato-dentritic [Cl\(^{-}\)]\(_i\) gradient in the presence (100%) or absence of NKCC1 in the AIS.

doi:10.1371/journal.pcbi.1002149.g006
frequency input to distal dendrites, the net mean current through GABA<sub>A</sub>R synapses switched from outward to inward whereas the same rate of input to the soma continued to produce strong outward currents (Fig. 7C). Thus, while increasing intraburst frequency can effectively enhance hyperpolarization in the soma, it rapidly becomes ineffective in dendrites and can even become depolarizing in distal dendrites. For a fixed intraburst frequency, \( E_{Cl} \) converged to different steady-state levels (Fig. 7D) with

---

**Figure 7. Dependency of \( Cl^- \) accumulation on the site of synaptic input and KCC2 level.** Trains (40 Hz) of inhibitory postsynaptic currents (IPSCs) at a synapse located at one of four positions: soma and proximal, middle, and distal dendrites (40, 100, and 240 \( \mu \)m from soma, respectively) in simulations without (A) and with (B) background synaptic input (\( f_{inh} = 0.4 \text{ Hz}, f_{exc} = 0.1 \text{ Hz} \)). For this set of simulations, a single dendrite was lengthened (and number of compartments increased to 60) relative to the cell geometry summarized in Fig. 1A. Inversion of the IPSC was evident in the distal dendrites under conditions without KCC2 (right panels). C. Mean intraburst IPSC became smaller (i.e. less hyperpolarizing) with increasing distance from the soma and with decreasing KCC2 level. Synaptic background activity was the same as in B. Mean IPSC was measured at a “test” synapse activated at 40 Hz for 200 ms every second over 50 s of simulated time. Steady state value of \( E_{Cl} \) (D) and rate at which \( E_{Cl} \) approaches steady state (E) for different KCC2 levels and distances of “test” synapse from the soma. Steady state \( E_{Cl} \) reported in D was measured as the value to which \( E_{Cl} \) converged when GABA<sub>A</sub>R at the test synapse were artificially held open. This convergence was fit with a single exponential to determine the time constant reported in E.

doi:10.1371/journal.pcbi.1002149.g007
different rates (Fig. 7E) depending on the location of the test synapse and the level of KCC2 activity. In other words, the steady-state value of [Cl\(^-\)], increased with distance from the soma (reminiscent of the standing gradient reported in Fig. 4A and C) and it decreased when KCC2 activity was increased. On the other hand, Cl\(^-\) accumulation converged to a steady state more rapidly with increased KCC2 activity as well as with distance from the soma. The two convergence processes are due to different phenomena: Enhanced KCC2 activity allows the dendrite to restrict the extent of Cl\(^-\) accumulation (see above), while Cl\(^-\) accumulates faster in distal dendrites simply because the effective volume is smaller and diffusion is restricted. In summary, under dynamic conditions, restricted diffusion in distal dendrites causes a rapid collapse of \(E_{\text{GABA}}\), but the extent of this collapse is limited by KCC2, consistent with experimental measurements [8,9,28].

In dendrites, distributed GABA\(_R\) input mediates greater inhibition than higher frequency focal input

The above results led us to predict that, for equivalent total synaptic input, many broadly distributed GABA\(_R\) synapses activated at low frequency would produce greater hyperpolarization than a few clustered synapses (or just one synapse) activated at higher frequency, especially for synapses located on distal dendrites. We tested this by comparing the outward current produced by one synapse activated at an intraburst frequency of 50 Hz with the total hyperpolarizing current produced by ten distributed synapses activated at 5 Hz; this was repeated for dendritically and somatically positioned synapses (Fig. 8A). In the soma, ten synapses activated at 5 Hz produced more outward current than one synapse activated at 50 Hz (Fig. 8A middle). This is due to the fact that the total synaptic conductance does not scale linearly with frequency because of saturation. Even more important is the fact that distributed dendritic input is capable of producing a strong outward current despite Cl\(^-\) accumulation, whereas clustered dendritic input was totally inefficient in producing an outward current. These results suggest that dendritic inhibition is most effective when spatially distributed, consistent with data in Figs. 3 and 6. Maintaining spatially distributed GABA\(_R\) synapses in dendrites is also important because the rapid dynamic collapse of distal hyperpolarizing GABA\(_R\) currents will limit their effectiveness at controlling somatic signals because membrane potential changes extend farther than changes in conductance [8,41]. Given that shunting remains even when \(E_{\text{GABA}}\) collapses, we submitted the neuron to distributed excitatory input and measured the mean firing frequency of the model neuron to verify that loss of hyperpolarizing current translates into effective disinhibition (Fig. 8A right). We found that firing rate reduction mirrored the change in charge carried (cf. Fig. 8A right and middle panels).

Enhancing GABA\(_R\) input may fail to enhance inhibition under conditions of impaired chloride homeostasis

In addition to synapse location, the rate and duration of synaptic inputs would be expected to interact with dynamic changes in \(E_{\text{GABA}}\) to alter the efficacy of inhibition. Although increasing the rate or duration of GABA\(_R\) inputs may initially increase IPSC amplitude, such changes would also accelerate depletion of the Cl\(^-\) gradient and thereby eventually reduce IPSC amplitude, at least when Cl\(^-\) influx overwhelms local diffusion mechanisms and Cl\(^-\) extrusion capacity. Using our model, we studied the influence of KCC2 activity level, synaptic frequency and time constant of GABA\(_R\)-mediated events (\(t_{\text{IPSC}}\)) on the mean current through a dendritic GABA\(_R\) synapse. Simulations indicated that increasing KCC2 activity always led to larger mean outward current. In contrast, increasing synaptic input frequency (Fig. 8B left) or \(t_{\text{IPSC}}\) (Fig. 8B right) did not necessarily increase the mean current; in both cases, the mean current was largest at intermediate values of those parameters. Similarly, mean firing rate was reduced most at intermediate values of those parameters (Fig. 8C). To establish the generality and robustness of the result, we repeated simulations for neurons endowed with different ion channels affecting spike generation. We added non-inactivating Ca\(^{2+}\)-activated K\(^+\) channels known to decrease firing rate or persistent Na\(^+\) channels known to increase firing rate, and we also performed simulations in which dendritic Hodgkin-Huxley (HH) channels were concentrated at branch points. Although these modifications to the model changed the overall firing rate, our qualitative finding remained unchanged; that is, firing rate increased if GABA\(_R\) input was augmented beyond a certain level (Fig. 8C right).

The above results indicate that more or longer GABA\(_R\) inputs may not always produce more inhibition, i.e. stronger outward current. We therefore asked what GABA\(_R\) input conditions produce the strongest inhibition? This question was addressed by measuring which parameter combinations produced the largest outward current. We found that the GABA\(_R\) input frequency yielding the largest outward current increased with KCC2 activity and decreased with \(t_{\text{IPSC}}\) (Fig. 8D). This optimal frequency was as low as 6 Hz when KCC2 activity was depleted to 10% of its normal value and \(t_{\text{IPSC}}\) was set to 50 ms; in other words, GABA\(_R\)-mediated synaptic events occurring either at lower or at higher frequencies than 6 Hz produced less outward current. The optimal GABA\(_R\) input frequency climbed to 28 Hz when KCC2 activity was set to baseline and \(t_{\text{IPSC}}\) was set to 10 ms. Thus, the optimal GABA\(_R\) input frequency may vary quite widely depending on other factors, but the key observation is that beyond some point (determined by the robustness of Cl\(^-\) homeostasis), more GABA\(_R\) input does not necessarily produce more inhibition. Increasing the frequency of GABA\(_R\) input showed a similar inverted bell-shaped curve when estimating effective inhibition with either total charge carried or firing rate reduction (Fig. 8B and C).

Net current through GABA\(_R\) depends on the balance of chloride and bicarbonate flux

Results of simulations presented in Figure 7 showed that the current through GABA\(_R\) could reverse polarity if there was sufficient accumulation of intracellular Cl\(^-\). However, as the Cl\(^-\) gradient collapses, one would expect Cl\(^-\) flux to stop, but not to change its direction; likewise, the IPSCs would be expected to become smaller but not to invert. Indeed, if the GABA\(_R\) is modeled as passing only Cl\(^-\) ions, the IPSC decreases in size as Cl\(^-\) accumulates intracellularly, but it does not reverse direction (Fig. 9A) thus showing that bicarbonate flux must be accounted for in order to explain IPSC inversion [42,43]. An important and novel feature of our model is that HCO\(_3^-\) is not assumed to be constant. Even if the relative stability of [HCO\(_3^-\)] has been shown to result from complex interaction between HCO\(_3^-\) efflux, carbonic anhydrase-mediated reaction and proton extrusion mechanisms, most models choose to consider it constant de facto. However, simulating the various mechanisms involved in [HCO\(_3^-\)] management proved a useful tool for investigating the legitimacy of assuming [HCO\(_3^-\)] is constant and for studying potential interactions between Cl\(^-\) and HCO\(_3^-\) dynamics. Bicarbonate efflux produces an inward current, but that current is (normally) masked by the larger outward current produced by Cl\(^-\) influx, since the permeability ratio between Cl\(^-\) and HCO\(_3^-\) anions is approximately 4:1 [2,43]. But as the Cl\(^-\)-mediated outward current becomes smaller, the HCO\(_3^-\)-mediated inward
current becomes relatively larger, eventually causing the net current through GABA\(_\text{A}\)R to become inward. Unlike the Cl\(^{-}\) gradient, the HCO\(_3\)\(^{-}\) gradient tends not to collapse (Fig. 9B) because intracellular HCO\(_3\)\(^{-}\) is replenished by carbonic anhydrase-catalyzed conversion of CO\(_2\), which can readily diffuse across the membrane \([44,45]\).

But although the reactants of the carbonic anhydrase-catalyzed reaction (i.e. CO\(_2\) and H\(_2\)O) are not depleted, the forward reaction produces H\(^{+}\) in addition to HCO\(_3\)\(^{-}\). By removing HCO\(_3\)\(^{-}\), GABA\(_\text{A}\)R activity would be expected to reduce the intracellular pH, which has been observed experimentally \([24]\). Since accumulation of intracellular H\(^{+}\) shifts the equilibrium point of

---

Figure 8. Efficacy of inhibition depends on spatial and temporal features of GABA\(_\text{A}\)R input. A. Schematic shows synapse positioning (left panel). GABA\(_\text{A}\)R input clustered at a single synapse (red) produced less outward current than the same total input distributed across ten spatially separated synapses (green), especially for input to the distal dendrites (center panel). To ensure that “total charge” translates into functionally relevant inhibition (i.e. reduction in spiking), we submitted the model to distributed excitatory input (\(f_{\text{exc}} = 0.2\) Hz) and measured firing rate. As expected, reduction in firing frequency was greater when inhibitory input was spatially distributed (right panel). B. Net charge carried through a “test” synapse (color) consistently decreased as KCC2 activity was reduced, but increasing the frequency (left panel), time constant (middle panel) or conductance (right panel) of input at that synapse did not necessarily increase current amplitude. For the left panel, the time constant was held at 10 ms while the input frequency and KCC2 level were varied; the dotted line shows optimal frequency, which is re-plotted in D. For the middle panel, the input frequency was held at 30 Hz while the time constant and KCC2 level were varied. For the right panel input frequency and time constant were held at 30 Hz and 10 ms respectively while the conductance and KCC2 level were varied. Background synaptic activity was included in these simulations (\(f_{\text{inh}} = 0.4\) Hz, \(f_{\text{exc}} = 0.1\) Hz). Test synapse was positioned at 50 \(\mu\)m from the soma. C. We performed simulations similar to that in B but added distributed excitatory input to assess inhibition on the basis of firing rate reduction rather than on the basis of total charge (left panel). The pattern of inverted bell-shaped curves is consistent with B, thus confirming a net change in inhibition at the whole cell level. The graph on the right illustrates results obtained from simulations with models including Ca\(^{2+}\)-activated K\(^{+}\) channels or persistent Na\(^{+}\) channels. We also concentrated dendritic HH channels at branch points while preserving the total conductance of these channels. Results were qualitatively the same as in the graph on the left. D. Optimal input frequency depending on KCC2 level and time constant (left panel) and the corresponding current (right panel). Black curves correspond to dotted line on left panel of B. Note that this is the optimal frequency for activation of a single “test” synapse; optimal input frequency would necessarily decrease as the number of activated synapses increased, although the exact relationship would depend on the spatial distribution of those active synapses (see A) as well as the level of background synaptic activity.

doi:10.1371/journal.pcbi.1002149.g008
the reaction, intracellular HCO$_3^-$ slowly decreases, with a time constant in the order of several seconds, which explains the small hyperpolarizing shift in $E_{\text{HCO}_3}$ seen in Figure 9B over long time scales. By $E_{\text{Cl}}$ and $E_{\text{HCO}_3}$ shifting in opposite directions, $E_{\text{GABA}}$ tends toward the membrane potential. We therefore predicted that reducing changes in $E_{\text{HCO}_3}$ would lead to greater changes in $E_{\text{Cl}}$.
and, vice versa, that reducing changes in $E_{Cl}$ would lead to greater
changes in $E_{HCO_3}$. To test the first prediction, $[HCO_3^-]$, was held
constant (thus maintaining HCO$_3^-$ efflux), which enhanced the
depolarizing shift in $E_{Cl}$ on the other hand, increasing
intracellular HCO$_3^-$ depletion by reducing proton extrusion via
the Na$^+$–H$^+$ exchanger (thus reducing HCO$_3^-$ efflux) mitigated
the depolarizing shift in $E_{Cl}$ (Fig. 9C). To test the second
prediction, $[Cl^-]$, was held artificially constant, which enhanced the
hyperpolarizing shift in $E_{HCO_3}$, conversely, increasing
intracellular Cl$^-$ accumulation by reducing Cl$^-$ extrusion via
KCC2 mitigated the hyperpolarizing shift in $E_{HCO_3}$ (Fig. 9D).
These results demonstrate a trade-off between stability of $[Cl^-]$ and
stability of intracellular pH based on their common reliance on
$[HCO_3^-]$. It remains an open question whether $[Cl^-]$, or
intracellular pH is more strongly regulated under normal
conditions, but one can reasonably extrapolate when KCC2
activity is reduced, that the primary depolarizing shift in $E_{Cl}$ will
conspire with a secondary hyperpolarizing shift in $E_{HCO_3}$ to
produce a large depolarizing shift in $E_{Cl}$ for $E_{Cl}$. This is particularly
relevant to steady state conditions because, on the time scale of
individual synaptic events, pH buffering mechanisms are not
saturated, while on longer time scales, the rate limiting
components of HCO$_3^-$ homeostasis are the slower kinetics of
the HCO$_3^-$ and H$^+$ membrane transporters.

The Cl$^-$/HCO$_3^-$ exchanger can also play a role in pH
management and Cl$^-$ homeostasis regulation. To gain some
insight into the impact of this exchanger, we repeated simulations of
Figure 9C–D adding different levels of Cl$^-$/HCO$_3^-$ exchanger
activity to the model. As is the case for such ion exchangers, the
Cl$^-$/HCO$_3^-$ exchanger will drive $E_{Cl}$ and $E_{HCO_3}$ towards one
another, namely depolarizing $E_{Cl}$ and hyperpolarizing $E_{HCO_3}$
(Fig. 9E). This result may seem counterintuitive since the
exchanger would be expected to play a helpful role in pH
management. However, in the instance of another source of
acidification, $E_{HCO_3}$ can undergo a hyperpolarizing shift, and the
resultant change in HCO$_3^-$ gradient can reverse Cl$^-$/HCO$_3^-$
transport, driving Cl$^-$ out and HCO$_3^-$ in, thus preventing overt
acidification (Fig. 9F).

These results predict that $E_{Cl}$ can become more hyperpolarized
during episodes of acidification. To test this, we modeled H$^+$ influx
occurring over 5 seconds and monitored the time course of $E_{Cl}$
during and after acidification in simulations with and without the
Cl$^-$/HCO$_3^-$ exchanger. In such simulations, proton influx triggers
a reaction with HCO$_3^-$ thus leading to a decrease in $[HCO_3^-]$. In
turn, this leads to hyperpolarization of $E_{HCO_3}$ which will eventually
become more hyperpolarized than $E_{Cl}$, effectively inverting the
exchanger and leading to hyperpolarization of $E_{Cl}$ (Fig. 9F). As
the influx of H$^+$ is stopped, H$^+$ extrusion through the Na$^+$–H$^+$ exchange restores pH and the carbonic anhydrase mediated reaction is able to
replenish intracellular HCO$_3^-$. As this slow change in $[HCO_3^-]$ trans-
lates into a change in the activity of the Cl$^-$/HCO$_3^-$
exchanger, the value of $E_{Cl}$ slowly becomes more depolarized until
it returns to its resting value (Fig. 9F). As expected, these changes in
$E_{Cl}$ cannot be observed when simulations are conducted without the
Cl$^-$/HCO$_3^-$ exchanger (Fig. 9F). Thus, the Cl$^-$/HCO$_3^-$
exchanger may be seen as a failsafe mechanism preventing overt
acidification, at least when this acidification is not caused by
HCO$_3^-$ efflux through GABA$_A$ channels.

**Accumulation of extracellular potassium influences GABA$_A$R-mediated current via a multi-step feedback loop**

To extrude Cl$^-$ from the cell, KCC2 must pass an equal number of K$^+$ ions since the net process is electroneutral. Therefore, K$^+$ efflux through KCC2 could reduce the transmem-
brane K$^+$ gradient and produce a depolarizing shift in $E_K$, which
would, in turn, reduce Cl$^-$ extrusion via KCC2 because of the
reduction in KCC2 driving force. To investigate this putative
negative feedback mechanism, we varied KCC2 activity and
measured the impact on $E_K$ (measured at the soma) in a model
neuron receiving a fixed level of background excitatory and
inhibitory synaptic input. Simulations showed that under condi-
tions of distributed GABA$_A$R input at in vivo-like background
frequencies, KCC2 activity actually had little impact on $E_K$ unlike
its large impact on $E_{Cl}$ (Fig. 10A, compare left and right panels). We
investigated this further by monitoring intra- and extracellular
concentrations of K$^+$ (Fig. 10B). Although large in absolute terms,
changes in K$^+$ were small in relative terms, yielding much smaller
shifts in $E_K$ than those observed with $E_{Cl}$. Furthermore, KCC2
activity had only a small influence on $[K^+]_i$, which is controlled
principally by the balance of K$^+$ leak conductance, active pumping
by the Na$^+$/K$^+$-ATPase, and extracellular diffusion.

The insignificant effect of KCC2 activity on $[K^+]_i$ is apparently inconsis-
tent with experimental observations [46], but those experiments
involved applying a heavy Cl$^-$ load, which is not comparable to the
physiological conditions tested in Figure 10A and B. To test whether a larger Cl$^-$ load could provoke a KCC2-
mediated increase in $[K^+]_i$, we simulated a constant 5 nS, 500
ms-long GABA$_A$R conductance on a dendrite. Under those
conditions, $[K^+]_i$ was significantly altered by KCC2 activity, as
shown by the positive correlation between the maximal value of
$[K^+]_i$, and KCC2 level (Fig. 10C). Repeating those simulations with
reduced extracellular K$^+$ clearance confirmed that extracel-
lular diffusion did not dramatically alter $[K^+]_i$ under these “heavy
load” conditions (Fig. 10C). Regardless of whether KCC2 activity
does or does not influence extracellular K$^+$ accumulation,
extracellular K$^+$ accumulation is nonetheless expected to reduce the
efficacy of KCC2 by reducing its driving force. To test this,
we repeated the simulations shown in Figure 1D with different fixed
values of $[K^+]_o$ and observed that the KCC2 efficacy is indeed
reduced by the extracellular K$^+$ accumulation and stops passing
ions when $[K^+]_o = 10$ mM (Fig. 10D).

It is important to understand that changes in $[K^+]_i$ have a much
larger effect on $E_K$ than equivalent absolute changes in
$[K^+]_o$. Hence, although KCC2 activity is not expected itself to change $E_K$
under normal physiological conditions (see above), changes in $E_K$
crushed by other factors (e.g. high firing rates, reduced Na$^+$/K$^+$-
ATPase activity, etc.) reduce KCC2 activity. In other words, there is
no closed negative feedback loop directly linking KCC2 and $E_K$,
but extrinsic factors can modulate Cl$^-$ extrusion by affecting extracel-
lular K$^+$ accumulation. Indeed, it is significant that Cl$^-$ extrusion
could be reduced (and inhibition thereby rendered ineffective)
under conditions where excessive spiking (perhaps the result of
disinhibition) causes extracellular K$^+$ accumulation – this would
constitute a multi-step positive feedback loop (see also below).

**Failure to control spiking increases chloride accumulation through a positive feedback loop that leads to catastrophic failure of inhibition**

As shown in previous sections, GABA$_A$R input and KCC2
activity are prominent determinants of $E_{Cl}$. However, since Cl$^-$
 influx depends on the Cl$^-$ driving force (i.e. $V – E_{Cl}$), variation
in membrane potential will influence intracellular Cl$^-$
accumulation, as shown in voltage clamp experiments [20]. Therefore, we
predicted that increased depolarization caused by increased
synaptic excitation would exacerbate intracellular Cl$^-$
accumulation. To test this, the frequency of inhibitory synaptic events, $f_{exc}$
was fixed at 0.4 Hz/synapse while the frequency of excitatory
synapses, $f_{exc}$, was varied (0.4 Hz was chosen for inhibitory events

---

**How Cl$^-$ Homeostasis Affects Synaptic Inhibition**
applying a 500 nS GABA conductance to a dendrite. Time constant for
diffusion from the FH space was tested at 100 and 200 ms (which
corresponds to normal and 50% slower extracellular K+ clearance) as
well as with variable extracellular space. D. $E_{CI}$ as a function of the mean
frequency of inhibitory input for various fixed levels of $[K^+]_o$.
doi:10.1371/journal.pcbi.1002149.g010

so that when $f_{exc}/f_{inh} = 2$, $f_{exc}$ was still within its normal
physiological range [24,30]. As predicted, the depolarizing shift in
$E_{CI}$ scaled with $f_{exc}$ (Fig. 11A). Moreover, given that spike
generation makes membrane potential a highly nonlinear function
of synaptic activity, we further predicted that the presence or
absence of spiking would have a profound influence on [Cl$^{-}$],
because each spike represents a large, albeit short, increase in Cl$^{-}$
driving force; in other words, if GABA$\text{A}R$ channels are open
during a spike, those spikes are expected to dramatically accelerate
intracellular Cl$^{-}$ accumulation. To test this, we measured Cl$^{-}$
accumulation in a model with and without spikes (i.e. with and
without HH channels, respectively). Results confirmed that Cl$^{-}$
accumulation was indeed increased by spiking (Fig. 11B). The time
series in Figure 11C shows the biphasic Cl$^{-}$ accumulation
associated with this phenomenon: When inhibition was first
“turned on”, it successfully prevented spiking but, over time,
[Cl$^{-}$]$_i$ increased asymptotically toward some steady-state value. If
the associated steady-state $E_{GABA}$ was above spiking threshold (as
in Fig. 11C), the membrane potential could increase beyond
threshold and the neuron began spiking, at which point
intracellular Cl$^{-}$ began a second phase of accumulation. This
second phase of Cl$^{-}$ accumulation was paralleled by acceleration
of the spike rate – clear evidence of the predicted positive feedback
loop between spiking and Cl$^{-}$ accumulation, which leads to
catastrophic failure of inhibition.

To verify experimentally the model prediction that excitatory
activity exacerbates intracellular Cl$^{-}$ accumulation, especially
when KCC2 activity is depleted, we performed [Cl$^{-}$]$_o$ measure-
ments in primary cultured neurons exposed to muscimol, followed
by addition of furosemide and kainate. The latter was to cause
tonic activation of AMPA subtype glutamate receptors. As
predicted by the model, addition of furosemide caused Cl$^{-}$
accumulation in the cell, and subsequent application of kainate led
to further accumulation (Fig. 11D).

The fact that $E_{CI}$ collapses as a result of GABA$\text{A}R$ activity itself
(Figs. 1, 3, 9) as well as excitatory input (Fig. 11A and D) and
spiking (Fig. 11B and C) highlights the importance of treating $E_{CI}$
as a dynamic variable. To assess the importance of those dynamics
on GABA$\text{A}R$ modulation of the firing rate, we compared the
relationship between firing rate and synaptic input in conditions
where both inhibitory and excitatory input change in a
proportional manner (i.e., $f_{inh} \propto f_{exc}$). We performed simulations
in which $E_{CI}$ was treated as a static value (as in conventional cable
models) or as a dynamic variable (as in our electrodiffusion model).
In the former case, $E_{GABA}$ was fixed at -65 mV, while in the latter
case, KCC2 activity was reduced to 33% of its normal level. With
weak excitatory and inhibitory input, spiking was higher in the
model with static $E_{CI}$ (Fig. 11E). However, as the frequencies of
excitatory and inhibitory inputs were increased, all the mecha-
nisms that contribute to a collapse of $E_{CI}$ (examined above)
combined to drive $f_{exc}$ nonlinearly beyond the value predicted by
fixed $E_{CI}$ (Fig. 11E). In short, these results show that $E_{CI}$ cannot be
approximated by a single, static value when considering a range of
stimulus conditions because of the rich dynamics governing $E_{CI}$
under natural conditions. Those dynamics can only be fully
understood by accounting for numerous, interdependent biophys-
ical processes.

Figure 10. Interactions between [Cl$^{-}$] regulation and [K$^+$]
regulation. A. Variation of KCC2 levels caused sizeable shifts in $E_{CI}$
(right panel) but had negligible effects on $E_K$ (left panel). Background
synaptic activity was $f_{exc} = 0.2$ Hz and $f_{inh} = 0.8$ Hz. B. Intra- and
extracellular concentrations of K$^+$ for same simulations reported in A.
Although extracellular K$^+$ levels are low, [K$^+$]$_o$ remains relatively stable
due to other mechanisms, e.g. extracellular diffusion. This explains why
$E_K$ remains relatively constant in A. C. Maximal [K$^+$]$_o$ reached by

How Cl- Homeostasis Affects Synaptic Inhibition

PLoS Computational Biology | www.ploscompbiol.org 14 September 2011 | Volume 7 | Issue 9 | e1002149
Figure 11. Effects of membrane potential on intracellular Cl\(^{-}\) accumulation. A. Varying the rate of excitatory synaptic drive (\(f_{\text{exc}}\)) caused a depolarizing shift in \(E_{\text{Cl}}\) secondary to changes in average membrane potential. \(f_{\text{inh}}\) was fixed at 0.4 Hz. B. Spiking exacerbates intracellular Cl\(^{-}\) accumulation as illustrated here by convergence of the model to different steady state \([\text{Cl}^{-}]_{i}\) depending on whether the model does or does not contain HH channels (i.e. does or does not spike, respectively). For this simulation, KCC2 activity was low (10%), \(f_{\text{inh}} = 0.8\) Hz, and \(f_{\text{exc}} = 0.4\) Hz. C. Sample traces showing inter-relationship between \([\text{Cl}^{-}]_{i}\) and spiking. Neuron began spiking when constant excitatory current was applied to the soma, but without any concomitant change in \([\text{Cl}^{-}]_{i}\) since there was not yet any GABA\(_{A}\)R-mediated conductance. Turning on constant GABA\(_{A}\)R conductance in the soma terminated spiking, but at the expense of intracellular Cl\(^{-}\) accumulation. Chloride slowly accumulated over the next several seconds until membrane potential reached spike threshold, at which point spiking resumed and Cl\(^{-}\) began a second phase of accelerated accumulation. D. To test whether Cl\(^{-}\) accumulation is exacerbated by excitatory synaptic input in real neurons, somatic Cl\(^{-}\) concentration was measured using FLIM in neurons with or without glutamatergic receptor activation by kainate. As predicted by simulations, Cl\(^{-}\) accumulation was greater in neurons exposed to kainate. Furosemide was applied to block KCC2 activity in these experiments (**, \(p < 0.001\); ***, \(p < 0.0001\)). Data from

---

**Figure 11. Effects of membrane potential on intracellular Cl\(^{-}\) accumulation.**

**A.** Varying the rate of excitatory synaptic drive (\(f_{\text{exc}}\)) caused a depolarizing shift in \(E_{\text{Cl}}\) secondary to changes in average membrane potential. \(f_{\text{inh}}\) was fixed at 0.4 Hz. **B.** Spiking exacerbates intracellular Cl\(^{-}\) accumulation as illustrated here by convergence of the model to different steady state \([\text{Cl}^{-}]_{i}\) depending on whether the model does or does not contain HH channels (i.e. does or does not spike, respectively). For this simulation, KCC2 activity was low (10%), \(f_{\text{inh}} = 0.8\) Hz, and \(f_{\text{exc}} = 0.4\) Hz. **C.** Sample traces showing inter-relationship between \([\text{Cl}^{-}]_{i}\) and spiking. Neuron began spiking when constant excitatory current was applied to the soma, but without any concomitant change in \([\text{Cl}^{-}]_{i}\) since there was not yet any GABA\(_{A}\)R-mediated conductance. Turning on constant GABA\(_{A}\)R conductance in the soma terminated spiking, but at the expense of intracellular Cl\(^{-}\) accumulation. Chloride slowly accumulated over the next several seconds until membrane potential reached spike threshold, at which point spiking resumed and Cl\(^{-}\) began a second phase of accelerated accumulation. **D.** To test whether Cl\(^{-}\) accumulation is exacerbated by excitatory synaptic input in real neurons, somatic Cl\(^{-}\) concentration was measured using FLIM in neurons with or without glutamatergic receptor activation by kainate. As predicted by simulations, Cl\(^{-}\) accumulation was greater in neurons exposed to kainate. Furosemide was applied to block KCC2 activity in these experiments (**, \(p < 0.001\); ***, \(p < 0.0001\)). Data from
Discussion

In this study, we built a neuron model that incorporates multiple processes controlling ion flux in order to investigate how interactions between those processes influence GABA$_A$R-mediated inhibition. This was prompted by the recognition that conventional neuron models make oversimplifying assumptions (e.g., reversal potentials are temporally invariant and spatially uniform or consider changes in only one ion species) that are likely to be particularly consequential for GABA$_A$R-mediated inhibition. For instance, experiments have shown that $E_{GABA}$ can shift during the course of sustained GABA$_A$R input [2,42], that $E_{GABA}$ is not uniform across different regions of the same neuron (our results and [26–28,46]) and that $E_K$ has an important impact on Cl$^-$ dynamics. Computational simulations are an ideal tool for investigating questions related to electrodiffusion and interaction between multiple ion species as well as for making predictions to guide subsequent experiments, but the accuracy of those simulations depends on the accuracy of the starting model. With that in mind, we built a neuron model that tracked [Cl$^-$] changes as well as other ions that interact with [Cl$^-$] homeostasis. Our model accurately reproduced activity-dependent decrease of IPSC amplitude, including differential decrease depending on the site of synaptic input and the compartment geometry [1,47]. Our model also reproduced spatial variations in $E_{GABA}$ and its dependence on the interplay between strength of cotransporter activity and spatial distribution of GABA$_A$R input. Having thus validated the model, we explored several other questions.

Upregulation of KCC2 has been linked with the hyperpolarizing shift in $E_{GABA}$ observed during early development [7,20,45,48]. Likewise, downregulation of KCC2 has been linked with the depolarizing shift in $E_{GABA}$ seen in various disease states [16,49,50]. However, the relationship between KCC2 and $E_{GABA}$ has not heretofore been quantitatively explored. Simulations in our electrodiffusion model showed that that relationship is highly nonlinear: Reducing KCC2 activity caused a dramatic depolarizing shift in $E_{GABA}$, whereas increasing KCC2 activity above normal levels had only a small effect on $E_{GABA}$. The reason is that KCC2 already operates near its equilibrium point under normal conditions [51]. These observations suggest that therapies aiming to restore depleted KCC2 levels should not cause excessively strong GABA$_A$R-mediated inhibition if KCC2 overshoots its normal level. Moreover, the importance of investigating KCC2 regulation as a therapeutic target is emphasized by the observation that increasing the frequency or duration of GABA$_A$R input cannot effectively compensate for disinhibition caused by KCC2 depletion since activity-dependent accumulation of intracellular Cl$^-$ is increased under those conditions. In fact, our simulations illustrate how the optimal rate and time course of GABA$_A$R input mutually influence each other and also depend on the level of KCC2 activity. Those observations help to explain why drugs that act by increasing GABA$_A$R input have variable effects on the treatment of pathological conditions involving disrupted Cl$^-$ homeostasis, e.g. in neuropathic pain or epilepsy. While administration of benzodiazepines has some efficacy at reversing tactile allodynia in neuropathic pain models, beyond a certain dose, they become counterproductive and enhance allodynia [32,53]. This bell-shaped response to benzodiazepines on neuropathic pain follows directly the predictions from our model (Fig. 8).

Beyond helping understand pathological conditions, our model also provides insight into synaptic inhibition under normal conditions. The importance of interactions between Cl$^-$ diffusion and transmembrane Cl$^-$ flux became apparent when we considered the temporal dynamics of [Cl$^-$]. Simulations revealed that Cl$^-$ accumulation near a highly active synapse is rapidly redistributed by intracellular diffusion, whereas Cl$^-$ extrusion via KCC2 tends to act more slowly. The large volume of the soma keeps somatic [Cl$^-$], relatively stable, in contrast to dendrites where diffusion is limited by the small diameter of the compartment. Thus, on short time scales, the soma acts as a Cl$^-$ sink. It follows that the extent of Cl$^-$ accumulation in dendrites does not only depend on the diameter of the dendrite, but also on the distance of the synapse from the soma. Since the dendrite diameter tends to decrease with the distance from the soma, the effects on diffusion are cumulative. As a result, Cl$^-$ diffusion is responsible for redistributing (and thus mitigating) transient, local changes in Cl$^-$ load, while KCC2 level controls the steady-state balance of Cl$^-$ influx and efflux. Thus, the faster dynamical collapse of $E_{GABA}$ that occurs upon repetitive GABA$_A$R input to distal dendrites results from limited diffusion rather than from inefficiency of Cl$^-$ extrusion.

The functional impact of this result is that distributed synaptic input is more effective than clustered input, especially on distal dendrites where longitudinal Cl$^-$ diffusion is particularly restricted. The more labile Cl$^-$ gradient in distal dendrites causes a rapid collapse of GABA$_A$R-mediated hyperpolarization upon repetitive input, which limits its ability to influence somatic integration especially because, although remote current sources can hyperpolarize the soma, remote conductances do not cause shunting in the soma [1]. This implies that multiple GABAergic connections originating from the same presynaptic cell will be more effective if those synapses are distributed on different dendritic branches. It is interesting to note that this corresponds to the morphological arrangement observed in several systems [54]. This broad distribution contrasts the clustering of axo-axonic synapses that necessarily occurs when a presynaptic cell forms multiple synapses on a postsynaptic neuron’s soma and AIS [54,55]. In the latter case, dynamical collapse of $E_{GABA}$ does not occur because the soma acts as a Cl$^-$ sink.

The functional impact of the standing [Cl$^-$], gradient along the somato-dendritic axis resulting from the interplay between background GABA$_A$R input and cotransporter activity may lead, under certain conditions, to differential impact of distal dendritic vs. somatic GABAergic synaptic input such as, for example, concurrent dendritic GABA$_A$-mediated excitation and somatic inhibition [1].

In addition to Cl$^-$ dynamics, one must keep in mind that Cl$^-$ flux does not occur independently from other ion species. For example, Cl$^-$ influx through GABA$_A$R is coupled with HCO$_3^-$ efflux. The relationship between Cl$^-$ flux and HCO$_3^-$ flux is crucial for explaining how the net current through GABA$_A$R can invert as Cl$^-$ accumulates intracellularly [2,44]. Beyond causing a given shift in $E_{GABA}$, the HCO$_3^-$ efflux has consequences on the dynamics of the system. Without HCO$_3^-$ efflux, Cl$^-$ influx would rapidly stabilize when membrane potential initially reached $E_{GABA}$ because $E_{GABA}$ would equal $E_{Cl^-}$. However, due to HCO$_3^-$ efflux, and given that $E_{GABA}$ is less negative than $E_{Cl^-}$ intracellular Cl$^-$ continues to accumulate when the membrane potential initially reaches $E_{GABA}$. In the absence of other extrinsic factors and during sustained GABA$_A$R input, intracellular Cl$^-$ accumulation and membrane potential drift would progress until $E_{Cl^-} = E_{GABA} = E_{HCO3}$. This progression may, however, be prevented by the
influence of other intrinsic currents. In any case, \( \text{HCO}_3^- \) efflux effectively delays stabilization of the system until a more depolarized membrane potential is reached, which can make a crucial difference for whether or not membrane potential increases above spike threshold (see below). Consistent with these observations, a recent study showed that blocking carbonic anhydrase (and thereby presumably reducing \( \text{HCO}_3^- \) efflux through \( \text{GABA}_A \text{R} \)) can mitigate some of the behavioral manifestations of neuropathic pain thought to arise from KCC2 downregulation \[52\]. Moreover, based on their common reliance on \( \text{HCO}_3^- \) regulation of \( [\text{Cl}]^- \), competes with regulation of intracellular \( \text{pH} \) on long time scales (tens of seconds to minutes) consistent with experimental observations \[3,24,56\]. One functional consequence of this is that intracellular \( \text{Cl}^- \) accumulation (and, by extension, possibly the loss of KCC2 expression in pathological conditions) may act as a protective mechanism to prevent an excessive drop in intracellular \( \text{pH} \) during sustained \( \text{GABA}_A \text{R} \) input.

The relationship between \( \text{pH} \) and \( \text{Cl}^- \) homeostasis may also be relevant to recent controversies regarding the necessity of ketone bodies for maintenance of \( E_{\text{GABA}} \) in the developing nervous system \[57–60\]. Given the \( \text{HCO}_3^- \) dependence of the beta-hydroxybutyrate effect on \( E_{\text{GABA}} \) in these experiments, it has been proposed that the explanation may reside in the fact that beta-hydroxybutyrate, lactate or pyruvate act as weak organic acids, thus acidifying the neuronal cytosol and reversing \( \text{Cl}^-/\text{HCO}_3^- \) exchange; this counteracts the drop in \( [\text{HCO}_3^-] \), due to acidification but, by the same token, lowers \( [\text{Cl}]^- \), and drives \( E_{\text{GABA}} \) to a more negative value \[59,61\]. Our simulations are consistent with this explanation.

Given the coupled efflux of \( \text{Cl}^- \) and \( K^+ \) through KCC2, \( \text{Cl}^- \) extrusion happens at the expense of extracellular \( K^+ \) accumulation. This may appear counter-productive as extracellular \( K^+ \) accumulation counteracts inhibition and plays a role in the onset of epilepsy \[62,63\]. However, we found that under physiological conditions, \( K^+ \) efflux through KCC2 is offset by the fact that KCC2 activity enhances inhibition, thus decreasing firing rate and reducing \( K^+ \) efflux via transmembrane channels. The net effect is a reduction of excitability because \( K^+ \) efflux via transmembrane channels is larger than via KCC2. We found that this negative feedback stabilizes \( [K^+]_o \) over a wide range of KCC2 activity. Disrupting this homeostasis requires sustained input from extrinsic factors. For example, intense \( \text{GABA}_A \)-ergic activity, which can maintain a continuous \( \text{Cl}^- \) load leading to a large and sustained \( K^+ \) efflux through KCC2, has been observed during giant depolarizing potentials \[46\]. Likewise, excessive spiking yields continuous extracellular \( K^+ \) accumulation, which renders KCC2 inefficient, causing a collapse of inhibition due to intracellular \( \text{Cl}^- \) accumulation.

Another interesting observation was that membrane depolarization tends to encourage intracellular \( \text{Cl}^- \) accumulation because \( \text{Cl}^- \) influx through \( \text{GABA}_A \text{R} \) depends on \( \text{Cl}^- \) driving force, which is increased by depolarization. The consequences are profound: If sustained \( \text{GABA}_A \text{R} \) input fails to prevent depolarization caused by concurrent excitatory input, the resulting depolarization will accelerate \( \text{Cl}^- \) influx, which in turn further reduces the \( \text{GABA}_A \text{R}-\)mediated outward current, thus supporting a positive-feedback cycle of failing inhibition. If the membrane potential reaches the spike threshold under these conditions, spike generation compounds the positive feedback process leading to an absolute failure of inhibition having potentially catastrophic consequences with respect to the neuron’s response to stimulation. The only way for a neuron to avoid entering this vicious cycle is to regulate \( [\text{Cl}]^-_o \) through \( \text{Cl}^- \) extrusion via KCC2.

In summary, we built a neuron model that incorporates multiple processes controlling the flux of different ion species in order to investigate how interactions between those processes influence inhibition mediated by \( \text{GABA}_A \text{R} \). Many of those processes cooperate or compete with one another, thus producing nonlinearities. The most dramatic of those is arguably the catastrophic failure of inhibition that can develop when depolarization and spiking conspire with \( \text{Cl}^- \) accumulation to form a positive feedback loop. As demonstrated in this study, such details may be critical for understanding important aspects of synaptic inhibition, in particular, for understanding why and how inhibition fails under certain pathological conditions.

**Methods**

We built a conductance-based model of a whole neuron using the NEURON simulation environment \[64\] (model code will be made available at ModelDB). The model is composed of 30 dendritic compartments unless otherwise indicated, one somatic compartment, one compartment for the axon initial segment (AIS), and 10 myelinated axonal compartments separated by nodes of Ranvier. Details of the geometry are summarized in Figure 1A. Ionic currents flowing through channels, pumps and cotransporters were computed at each time step in order to update the membrane potential according to \( \frac{dV}{dt} = -\sum I \) where \( C \) is membrane capacitance of the neuron compartment and the sum is taken over synaptic currents, current through voltage gated channels and electrogenic Na-K ATPase. Transmembrane ion flux due to those currents was also calculated. Moreover, longitudinal and radial diffusion were incorporated into the model in order to account for intracellular ion gradients. Likewise, extracellular ionic diffusion was taken into account as well as chemical reactions that produce the various ion types (see below). Transmembrane ion flux, ion movement through diffusion, and ion generation through chemical reactions (Fig. 1B) were all taken into account when updating the concentration of ion species \( x \) in each compartment at each time step according to the differential equation \( \frac{d[x]_o}{dt} = \pm I_x \cdot (z \cdot F \cdot V_o) + \text{Diff}_x \cdot \frac{V_o}{V_t} + \text{Reac}_x \cdot \frac{V_o}{V_t} \), where \( F \) is the Faraday constant, \( z \) is the ion valence, \( V_o \) is the compartment volume, \( \text{Reac}_x \) is a term accounting for chemical reactions involving ion species \( x \) and \( \text{Diff}_x \) is a term modeling the electrodiffusion of ion \( x \) \[65,66\]. Synaptic events are expressed in currents, but the membrane potential was not clamped, consistent with realistic conditions. This is of importance since invasive cell manipulations have been shown to alter the nature (inhibitory or excitatory) of \( \text{GABA}_A \) mediated input \[67\].

**Channels**

Ions currents obey the equation \( I_x = g_x \cdot (V - E_x) \) where \( E_x \) denotes the reversal potential for ion \( x \) and \( g_x \) is the channel conductance with respect to ion \( x \). Reversal potentials were continuously updated during the simulation using the Nernst equation \( E_x = z \cdot R \cdot T \cdot \log \left( \frac{[x]}{[x]_o} \right) \) where \( R \) is the perfect gas constant and \( T \) is absolute temperature, which was taken to be 310 K (37 °C). Because \( \text{GABA}_A \) receptors pass both \( \text{Cl}^- \) and \( \text{HCO}_3^- \) anions in a 4:1 ratio \[24\], \( E_{\text{GABA}} \) was calculated using the Goldman-Hodgkin-Katz equation

\[
E_{\text{GABA}} = -\left( \frac{RT}{F} \right) \log \left( \frac{4 [\text{Cl}]^-_o + [\text{HCO}_3^-]_o}{4 [\text{Cl}]^-_o + [\text{HCO}_3^-]_o} \right)
\]

Each of these ionic currents was taken into account for computing change in concentration of their respective ion species.
and their sum yielded the net current used to update the membrane potential.

Synaptic input was modeled as a Poisson process. Each inhibitory synapse was activated at a mean frequency of 0–10 Hz and each excitatory synapse was activated at a mean frequency of 0–2 Hz. Unless otherwise stated, the maximal conductance of inhibitory synapses was 1 ± 0.3 nS (mean ± standard deviation) and kinetics were modeled as instantaneous rise and exponential decay with t_{\text{psc}} of 30 ms [30,60–70]. GABA\textsubscript{A}/R synaptic density was 60 synapses per 100 μm\textsuperscript{2} in the AIS, 40 synapses per 100 μm\textsuperscript{2} in the soma and 12 synapses per 100 μm\textsuperscript{2} in the dendrites. Density of excitatory synapses was 60 synapses per 100 μm\textsuperscript{2} in dendrites and no excitatory synapses were present elsewhere [21,30]. Unless otherwise stated, the maximal conductance of excitatory synapses was taken to be 0.5 ± 0.2 nS (mean ± standard deviation) and the kinetics were modeled as an instantaneous rise and exponential decay with t_{\text{psc}} of 10 ms.

Hodgkin-Huxley (HH) channels were modeled using parameter values reported by [12]. The voltage-dependant Na\textsuperscript{+} current was given by

\[
I_{Na} = g_{Na}m^4h(V - E_{Na})
\]

\[
\frac{dm}{dt} = z_m(V)(1 - m) - \beta_m(V)m
\]

\[
\frac{dh}{dt} = z_h(V)(1 - h) - \beta_h(V)h
\]

\[
z_m = -0.32(V - V_T - 13) \exp[(V - V_T - 40)/5] - 1
\]

\[
z_h = 0.128 \exp[ - (V - V_T - 17)/18]
\]

\[
\beta_m = 0.28(V - V_T - 40) \exp[(V - V_T - 40)/5] - 1
\]

\[
\beta_h = \frac{4}{1 + \exp[ - (V - V_T - 17)/18]}
\]

Where \(V_T = -58\) mV and \(V_s = -10\) mV. The voltage gated K\textsuperscript{+} channels were described by

\[
I_K = g_Kn^4(V - E_K)
\]

\[
\frac{dn}{dt} = z_n(V)(1 - n) - \beta_n(V)n
\]

\[
z_n = -0.032(V - V_T - 15) \exp[ -(V - V_T - 15)/5] - 1
\]

\[
\beta_n = 0.5 \exp[ -(V - V_T - 10)/40]
\]

The density of HH channels was 12 mS/cm\textsuperscript{2} in the AIS and 1.2 mS/cm\textsuperscript{2} in soma and dendrites [21,30]. The model also included K\textsuperscript{+} and Na\textsuperscript{+} leak channels with respective densities of 0.02 mS/cm\textsuperscript{2} and 0.004 mS/cm\textsuperscript{2} in soma, 0.03 mS/cm\textsuperscript{2} and 0.006 mS/cm\textsuperscript{2} in proximal dendrites, 0.1 mS/cm\textsuperscript{2} and 0.02 mS/cm\textsuperscript{2} in distal dendrites, 0.02 μS/cm\textsuperscript{2} and 0.004 μS/cm\textsuperscript{2} in axon internodes, and 15 mS/cm\textsuperscript{2} and 3 mS/cm\textsuperscript{2} in axon nodes as described in [21,30].

For some simulations, we added other types of conductances to account for the many possible types of spike generating mechanisms. Namely, we added non-inactivating Ca\textsuperscript{2+}-activated K\textsuperscript{+} channels and persistent Na\textsuperscript{+} channels to test spike reducing and spike enhancing mechanisms, respectively. The Ca\textsuperscript{2+}-activated K\textsuperscript{+} channels obey the following sets of equations

\[
I_{K_{Ca}} = g_{K_{Ca}}m^4h(V - E_{K_{Ca}})
\]

\[
\frac{dm}{dt} = \frac{m_0 - m}{\tau_m},
\]

\[
\frac{dz}{dt} = \frac{z_0 - z}{\tau_z},
\]

\[
\frac{dh}{dt} = \frac{h_0 - h}{\tau_h}.
\]

Where the auxiliary functions are defined by

\[
m_x = \frac{1}{1 + \exp(-(V - c_{xmn})/c_{xkm})}
\]

\[
\tau_m = \frac{1}{\exp(-(V - c_{xmn})/c_{xkm}) + \exp(-(V - c_{xmn2})/c_{xkm2})}
\]

\[
z_x = \frac{1}{1 + \exp( - (V + c_{xch})(c_{xch}) - c_{xch})}
\]

\[
h_x = \frac{1}{1 + \exp(-(V + c_{xch1})/c_{xch1}) + \exp(-(V + c_{xch2})/c_{xch2})}
\]

With the constants defined as \(c_{xmn} = 29.9\) mV, \(c_{xkm} = 6.2\) mV, \(c_{xkm1} = 0.000505\) s, \(c_{xkm1} = 86.4\) mV, \(c_{xmn2} = -10.1\) mV, \(c_{xkm2} = -33.3\) mV, \(c_{xch} = 10\) mV, \(c_{xch} = 0.085\), \(c_{xch} = 32\) mV, \(c_{xch} = 5.8\) mV, \(c_{xch} = 0.0019 s, c_{xch1} = 48.5\) mV, \(c_{xch2} = -54.2\) mV, \(c_{xch2} = 12.9\) mV.

The persistent Na\textsuperscript{+} channels were described by the following set of equations:

\[
I_{Na,p} = g_{Na,p}mhb(V - E_{Na}),
\]

\[
\frac{dm}{dt} = \frac{m_x - m}{\tau_m},
\]

\[
\frac{dh}{dt} = \frac{h_x - h}{\tau_h}.
\]

Where the auxiliary functions are defined by

\[
\tau_m = \left( \frac{0.32(v_{mn} + 13 - V)}{\exp\left(\frac{v_{mn} + 13 - V}{4}\right) - 1} + \frac{0.28(v_{mn} + V - 40)}{\exp\left(\frac{v_{mn} + V - 40}{5}\right) - 1} \right)^{-1}
\]

\[
m_x = \left( \frac{0.32(v_{mn} + 13 - V)}{\exp\left(\frac{v_{mn} + 13 - V}{4}\right) - 1} \right)^{-1}
\]

\[
\tau_h = \left( \frac{0.32(v_{mn} + 13 - V)}{\exp\left(\frac{v_{mn} + 13 - V}{4}\right) - 1} + \frac{0.28(v_{mn} + V - 40)}{\exp\left(\frac{v_{mn} + V - 40}{5}\right) - 1} \right)^{-1}
\]

\[
h_x = \frac{0.128 \exp((v_{ch} + 17 - V)/18)}{1 + \exp((v_{ch} + 40 - V)/2.5)}
\]

\[
h_x = \frac{0.128 \exp((v_{ch} + 17 - V)/18)}{1 + \exp((v_{ch} + 40 - V)/2.5)}
\]
The maximal $\text{Cl}^-$ current going through KCC2 was taken to be $0.003 \, \text{mS/cm}^2$ for the normal activity level which was taken to equal the value obtained for maximal current through KCC2. Na$^+$ and K$^+$ currents through NKCC1 were each half of $I_{\text{Cl},\text{NKCC1}}$ so that net current through NKCC1 was equal to zero.

Finally, in some simulations we also modeled the $\text{Cl}^-$/$\text{HCO}_3^-$ exchanger which was assumed to be ubiquitous in the neuron and uniformly distributed on the membrane. Kinetics of the exchanger were described by the following simplified equation.

$$I_{\text{Cl,exch}} = \frac{I_{\text{Cl},\text{max}}(E_{\text{HCO}_3} - E_{\text{Cl}})}{(E_{\text{HCO}_3} - E_{\text{Cl}}) + V_{\text{half}}}$$

where $I_{\text{Cl},\text{max}} = 0.1 \, \text{mA/cm}^2$ and $V_{\text{half}}$ was set to 50 mV. The $\text{HCO}_3^-$ and $\text{Cl}^-$ currents through the exchanger were taken to be equal in amplitude but opposite in direction so that the exchange process is electroneutral.

### Bicarbonate and pH management

Since $\text{HCO}_3^-$ ions also flow through GABA$\text{A}$R channels (see above), it was important to model the ionic fluxes and reactions regulating [HCO$_3^-$]. Intracellular $\text{HCO}_3^-$ loss due to outgoing flux via GABA$\text{A}$R is compensated by the carbonic anhydrase-catalyzed reaction $\text{CO}_2 + \text{H}_2\text{O} \rightarrow \text{H}^+ + \text{HCO}_3^-$. Since the diffusion of $\text{CO}_2$ gas through the membrane is faster than ionic fluxes through channels, we treated $\text{pCO}_2$ as constant. The equilibrium constant of the reaction was $10^{-6.35}$ M and the rate constant for $\text{CO}_2$ hydration was $10^{5}$ sec$^{-1}$ [75]. Since the above reaction produces a drop in pH [42,44,56] by causing intracellular $\text{H}^+$ accumulation, we also modeled the reaction between $\text{H}^+$ and the main buffering ion $\text{H}_2\text{PO}_4^-$ such that $\text{H}^+$ buffering occurred through the reaction $\text{H}^+ + \text{H}_2\text{PO}_4^- \rightarrow \text{H}_3\text{PO}_4^-$. Although other reactions play important roles in pH buffering, we kept the model as simple as possible while preserving the global value of pH buffering capacity, estimated to be 25–30 mEq/mol/pHU [76,77]. Buffering reactions are responsible for maintaining the pH value constant at short time scales (<100 ms), but proton extrusion via exchangers plays an important role on longer time scales (>10s). For the sake of simplicity, we limited ourselves to modeling the Na$^+$-H$^+$ exchanger such that the proton flux was given by

$$I_{\text{H,NaH}} = I_{\text{H, max}} \frac{(E_{\text{Na}} - E_{\text{H}})}{(E_{\text{Na}} - E_{\text{H}}) + V_{\text{half}}}$$

which is a generic scheme for transporters. We used $I_{\text{H, max}} = 0.03 \, \text{mA/cm}^2$ and $V_{\text{half}} = 10 \, \text{mV}$. These values were chosen to make the model consistent with the global proton extrusion rate in healthy neurons that has been measured to be 0.04 pHU/s [78–80].

### Electrodiffusion

An important and novel feature of the model is that the intra- and extracellular concentrations of K$^+$, Na$^+$, Ca$^{2+}$, Cl$^-$, HCO$_3^-$ as well as intracellular concentrations of H$^+$, HPO$_4^{2-}$ and H$_2$PO$_4^-$ were treated as dynamical variables updated in each compartment at each time step. Each compartment was divided in four concentric annulus-shaped subcompartments to account for radial diffusion. Diffusion coefficients were assumed to be the same as in water ($10^{-5}$ cm$^2$/s): 2.05 for Cl$^-$, 1.33 for Na$^+$, 1.96 for K$^+$, and 9.33 for H$^+$ [81].

The longitudinal electrodiffusion is described by the equation

$$\frac{\partial [x]}{\partial t} = D_x \frac{[x]}{\varepsilon^2} + \frac{D_y}{\varepsilon} \frac{\partial}{\partial y} \left( \frac{[x]}{\varepsilon} \right) + \frac{D_z}{\varepsilon} \frac{\partial}{\partial z} \left( \frac{[x]}{\varepsilon} \right)$$

where $y$ stands for the longitudinal axis, $D_x$ for the diffusion coefficient with respect to ion specie $x$, $Vol$ for section volume and $Surf$ for the surface of the cross section [66].
The first term is due to pure diffusion while the second term accounts for the electrical force acting on the ions. The second term was used only to compute electrodiffusion between outer annuli of dendritic sections and was set to 0 for inner annuli, consistent with the fact that electrical field extends only to a thin region near the membrane. This is because the membrane acts as a capacitor and electric field is known to decrease rapidly [82]. Radial diffusion was computed in a similar way but with y representing the radial axis.

Extracellular space was represented as a thin shell (i.e. Frankenhaeuser-Hodgkin or FH space) with equivalent volume equal to one fourth the intracellular volume of the corresponding cell compartment. The inner surface of the FH space communicated with the adjacent intracellular compartment while the outer surface was linked to an infinite reservoir where ion concentrations were assumed to be constant. This modeling takes into account changes in [K+]o due to our cell, and thus does not address changes in [K+]i due to network activity. The study of such network related effects is beyond the scope of the current study. The equation used to update extracellular ion concentration is given by

\[
\frac{d[X]}{dt} = -\frac{I_e}{(z\cdot F \cdot Vol)} + \frac{Diff_z [X] - k_{\text{radial}}}{\tau_{FH}}
\]

where \( z \) is the ion valence, \( k_{\text{radial}} \) is the concentration of ion \( x \) in the infinite reservoir and \( \tau_{FH} \) is the time constant taken to be 100 ms [79,83].

### Numerical methods

The differential equations were integrated using a forward Euler method with a time step of 0.05 ms. Several preliminary simulations showed this time step to be both sufficiently small for accurate equation solving, while sufficiently large for reasonably fast computing. Initial intracellular concentrations were ([mM]: [Cl-]i = 6, [K+]i = 140, [Na+]i = 10, [HCO3-]i = 15, [H2PO4-]i = 30 and [HPO42-]i = 30. Initial extracellular concentrations were ([mM]: [Cl-]o = 120, [K+]o = 3, [Na+]o = 45 and [HCO3-]o = 25 [81]. Preliminary simulations were conducted to determine initial concentrations such that they were stable under normal conditions (in the absence of high frequency synaptic input). For simulations in which the value of maximal Cl- current through KCC2 (\( I_{\text{Cl, KCC2}} \)) was different than the normal one stated above (\( I_{\text{Cl, base}} = 0.3 \text{mA/cm}^2 \)), different initial values of \([\text{Cl}^-]_i\) were used in order to start the simulation near steady state, as determined by preliminary testing. Unless stated otherwise, simulations were run for 200 s of simulated time, short enough to allow manageable simulations and long enough to allow collection of sufficiently large data sample to insure relevance of mean values.

### Cell cultures

Dissociated hippocampal neurons from Sprague-Dawley rats were prepared as previously described [84] plated at P0 to P2 at a density of approximately 500–600 cells/mm² and imaged after 21–30 days in vitro (DIV). Glial proliferation was stopped at 5 DIV with Ara-C.

### Chloride imaging

Cells were loaded in a 5 mM solution of the Cl- indicator MQAE (6-methoxyquinolinium acetoxylester; Molecular Probes) for 30 min at 37 °C [85]. Prior to observation, cells were transferred to a perfusion chamber and bathed in bicarbonate-buffered saline containing: 100 NaCl, 2.5 KCl, 1 NaH2PO4, 26 NaHCO3, 1 MgCl2 and 1.2 CaCl2 Muscimol (100 mM, Tocris), furosemide (50-200 μM, Sigma), kainic acid (50 nM, Tocris), VU 0240551 (25-50 μM, Tocris) and bicuculline (100 μM, Sigma) were selectively added as described in the result section. Upon addition of drugs, cells were allowed to adjust for 10–20 minutes before a steady-state image of their Cl- contents was taken.

Fluorescence lifetime images of MQAE were acquired using a Becker & Hickl SPC-830 module coupled to a Zeiss LSM 510 microscope. MQAE was excited using a femtosecond pulsed Ti:Sapphire laser tuned at 760 nm (Chameleon Ultra, Coherent), through a 40X water-immersion objective (Zeiss, 0.8 NA). Fluorescence lifetime data was collected through the non-descanned port of the microscope using a band-pass filter (469/55 nm, Semrock) coupled to a laser block (short-pass 750 nm; Semrock). Photon emission was detected using a PM100-1 photosensor (Hamamatsu). Lifetime in each cell compartment was calculated and extracted using SPCImage software (Becker & Hickl). Lifetime in the cell body was averaged over the total cell body area excluding the nucleus region, whereas in the dendrites it was averaged in segments of 4 μm over 120 μm of dendrite length. Fluorescence lifetime measurements were used because they are not sensitive to dye concentration (peak intensity) in the range we are using [85,86]. The lifetime measurements are thus not affected by differences in dye loading from cell to cell or by volume changes that could occur in different cell compartments (Fig. 2A). The Cl- dependence of MQAE lifetime is described by the Stern-Volmer relation (\( \tau_o/\tau = 1 + K_{sv} [\text{Cl}^-]_o \)), where \( \tau_o \) is the fluorescence lifetime in 0 mM Cl- and \( K_{sv} \), the Stern-Volmer constant, is a measure of the Cl- sensitivity of MQAE (Fig. 2B).

### Chloride calibration in cells

For calibration of absolute Cl- concentrations, the fluorescence lifetime of MQAE-loaded cells was measured in the presence of different known extracellular [Cl-] (8, 15 or 20 mM) in the bath. To dissipate the Cl- gradient across the membranes, 20 μM tributyltin (Cl-:OH exchanger) was used and 20 μM nigericin (K+–H+ exchanger) was added to clamp the intracellular pH using high K+ driving force while Cl- changes. Calibration solutions contained (in mM) KCl and KNO3 (140 K+ total with desired amount Cl-), 10 D-glucose, 10 HEPES, 1.2 CaCl2, 1 MgCl2, pH adjusted to 7.2 using KOH.

### Immunocytochemistry

Primary hippocampal cultures (5 and 28 DIV) were fixed for 10 min with 0.2% paraformaldehyde and then permeabilized for 45 min with 0.2% triton in 10% normal goat serum (NGS). Primary antibody incubations were performed overnight at 4 °C using a monoclonal marker of KCC2 (Rabbit anti-KCC2 1:500, Upstate) in the presence of 5% NGS. Alexa 546 conjugated secondary antibodies (1:750; Invitrogen, Eugene, OR) were applied for 2 hrs at room temperature. Images were obtained on the Zeiss LSM 510 microscope using a 63X/1.4NA oil objective (Zeiss).

### Spatial intensity distribution analysis (SpIDA)

SpIDA is a recently developed analysis method that can resolve concentration of mixtures of different monomeric and oligomeric labels in single fluorescence images by fitting its intensity histogram. Precise details of the technique and detector calibration are presented in [33]. Briefly, the intensity histogram fitting function for a system with density of \( N \) particles is:

\[
H(x;N;k) = \sum_{n} \rho^n(x;k) \cdot \text{pois}(n, N) \cdot \delta(k - n)
\]

Where \( \rho^n(x;k) = \rho^1(x;k) \cdot \rho^{n-1}(x;k) \) with \( \rho^1(x;k) = \frac{\delta(x;I(x))}{k!} \) is the illumination intensity profile of the excitation laser, \( \delta \) represents the quantal brightness of a single fluorescent
particle, and $k$ is the probability of observing an intensity of light (assumed to be proportional the number of photons emitted). $H$ is normalized over all intensity values so the integral over $k$ gives one. A constant factor, $A$, is introduced, which is the number of pixels in an analyzed region of the image where the fluorescent particles are distributed. This allows for the fit of an image intensity histogram to be performed. Three parameters are fit: the number of pixels [$A$], the fluorescent particle density [$\lambda$ particles per laser beam effective focal volume] and the quantum brightness ($\lambda$ intensity units, $\mu$ per unit of pixel integration time). In confocal laser scanning microscopes, the fluorescence intensity is measured using photon multiplier tubes (PMTs), and the number of collected photoclectrons is a function of the polarization voltage.

If dimers are present in the sample, they will yield quanital brightness of $2e$. When the monomer and dimer populations are mixed within the same region in space, the total histogram becomes the convolution of the two individual distributions:

$$H_{(e_1,N_1,e_2,N_2,k)} = A \cdot H_{(e_1,N_1,k)} \otimes H_{(e_2,N_2,k)}$$

To obtain accurate results, noise characteristics of the detector also has to be studied and taken into account in the analysis. See [33] for complete analysis.

For each sample, an optimal setting of the laser power and PMT voltage was chosen to minimize pixel saturation and photobleaching. The CLSM settings were kept constant for all samples and controls (Laser power, filters, dichroic mirrors, polarization voltage, scan speed). Acquisition parameters were always set within the linear range of the detector which was determined by calibration [33]. All the images were $1024 \times 1024$ pixels with pixel size of 0.115 $\mu$m and 9.1 $\mu$m pixel dwell time. The z-stacks were taken by optical sectioning with a $z$ step of 0.5 $\mu$m per image.

**Ethics statement**

All experiments were performed in accordance with regulations of the Canadian Council on Animal Care.

**Author Contributions**

Conceived and designed the experiments: ND SAP AC AGG YDK. Performed the experiments: AC. Analyzed the data: ND AC AGG. Contributed reagents/materials/analysis tools: ND AGG. Wrote the paper: ND SAP AC AGG HR YDK.

**References**

1. Hausser M, Major G, Stuart GJ (2001) Differential shunting of EPSPs by action potentials. Science 291: 138–141.
2. Staley KJ, S Hold BL, Proctor WR (1995) Ionic mechanisms of neuronal excitation by inhibitory GABA receptors. Science 269: 977–981.
3. Kaila K, Voipio J (1987) Postsynaptic Fall in intracellular pH induced by GABA-activated bicarboxionate Conductance. Nature 330: 163–165.
4. Staley KJ, Proctor WR (1999) Modulation of mammalian dentritic GABA receptor function by the kinetics of CO$_2$ and HCO$_3^-$ transport. J Physiol 519: 693–712.
5. Prescott SA, Sejnowski TJ, De Koninc K (2006) Reduction of axon reversal potential subverts the inhibitory control of firing rate in spinal lamina I neurons: towards a biophysical basis for neuropathic pain. Mol Pain 2: 32.
6. Blaesse P, Airaksinen MS, Rivera C, Kaila K (2009) Cation-Chloride Cotransporters and Neuronal Function. Neuron 61: 820–830.
7. Rivera C, Voipio J, Payne JA, Rauschecker E, Lah twin H, et al. (1999) The K$^+$/Cl$^-$ co-transporter KCC2 renders GABA hyperpolarizing during neuronal maturation. Nature 397: 251–255.
8. Brumbach AC, Staley KJ (2000) Thermodynamic regulation of KCC2-mediated Cl$^-$ co-transport undrules plasticity of GABA$_A$ signaling in neurons. J Neurosci 20: 1301–1312.
9. Coull JA, Boudreau D, Bachand K, Prescott SA, Nault F, et al. (2003) Transynaptic shift in anion gradient in spinal lamina I neurons as a mechanism of neuropathic pain. Nature 424: 938–940.
10. Coull JA, Boudreau D, Bouchard D, Boivin D, Tsuda M, et al. (2005) BDNF from microglia causes the shift in neuronal anion gradient underlying neuropathic pain. Nature 430: 1017–1021.
11. Dahal VI, Staley KJ (2003) Excitatory actions of endogenously released GABA contribute to initiation of ictal epileptiform activity in the developing hippocampus. J Neurosci 23: 1840–1846.
12. Glykys J, Staley KJ (2008) Differences in cortical vs. subcortical GABA effects: A mechanism for electrochemical dissociation of neonatal seizures. Epilepsia 49: 170.
13. Huberfeld G, Wittner I, Glennencau S, Baulac M, Kaila K, et al. (2007) Perturbed chloride homeostasis and GABAergic signaling in human temporal lobe epilepsy. J Neurosci 27: 9086–9087.
14. Kaila K, Staley KJ (2008) The bumetanide-sensitive Na-K-2Cl cotransporter NKCC1 as a potential target of a novel mechanism-based treatment strategy for neonatal seizures. Neuroscience 148: 938–940.
15. Payne JA, Rivera C, Voipio J, Kaila K (2005) Cation-chloride co-transporters in neuronal communication, development and trauma. Trends Neurosci 26: 199–206.
16. De Koninc K (2007) Altered chloride homeostasis in neurological disorders: a new target.Curr Opin Pharmacol 7: 93–99.
17. Woodin MA, Ganguly K, Poo MM (2003) Coincident pre- and postsynaptic activity modifies GABAergic synapses by postsynaptic changes in Cl$^-$ transporter activity. Neuron 39: 807–820.
18. Chabwine JN, Van Damme P, Eggermont J, De Smedt H, Missiaen L, et al. (2004) Long-lasting changes in GABA responsiveness in cultured neurons. J Physiol 569: 67–92.
19. Jedrika P, Deller T, Gutka BS, Backus K (2010) Activity-dependent intracellular chloride accumulation and diffusion controls GABA receptor-mediated synaptic transmission. Hippocampus 29: 509–519.
20. Cordero-Erausquin M, Coull JA, Boudreau D, Rolland M, De Koninc K (2005) Differential action and axon reversal potential in spinal lamina I neurons: impact of chloride exclusion capacity. J Neurosci 25: 9613–9623.
21. Devede A, Rudolph M, Pare D (2003) The high-conductance state of neocortical neurons in vivo. Nat Rev Neurosci 4: 739–751.
22. Rivera C, Voipio J, Thomas-Crusell J, Li H, Emri Z, et al. (2004) Mechanism of activity-dependent down-regulation of the neuron-specific KCl cotransporter KCC2. J Neurosci 24: 4683–4691.
23. Rivera C, Li H, Thomas-Crusell J, Lahitbin H, Vittanen T, et al. (2002) BDNF-induced TriB activation down-regulates the K$^+$-Cl$^-$ cotransporter KCC2 and impairs neuronal Cl$^-$ exclusion. J Cell Biol 159: 747–752.
24. Voipio J, Parnernack M, Rydpduit B, Kaila K (1991) Effect of Gamma-Aminobutyric-Acid on Intracellular pH in the Grayish Stretch-Receptor Neuron. J Exp Biol 134: 78–91.
25. Delpire E, Days E, Lewis LM, Mi D, Kalm K, et al. (2009) Small-molecule screen identifies inhibitors of the neuronal KCl cotransporter KCC2. Proc Natl Acad Sci U S A 106: 5383–5388.
26. Berglund K, Schleiw W, Krieger P, Luo LS, Wang DU, et al. (2006) Imaging synaptic inhibition in transgenic mice expressing the chloride indicator, Clomeleon. Brain Cell Biol 35: 207–228.
27. Berglund K, Schleiw W, Wang H, Feng GP, Hall WC, et al. (2008) Imaging synaptic inhibition throughout the brain via genetically targeted Clomeleon. Brain Cell Biol 36: 101–118.
28. Khiroug S, Yamada J, Azafalv R, Voipio J, Khiroug L, et al. (2008) GABAergic depolarization of the axon initial segment in cortical principal neurons is caused by the Na-K-Cl cotransporter NKCC1. J Neurosci 28: 4635–4639.
29. Durbel J, Haverkamp S, Schleiw W, Feng GP, Augustine GJ, et al. (2006) Two-photon imaging reveals somatodendritic chloride gradient in retinal ON-type bipolar cells expressing the biosensor Clomeleon. Neuron 49: 81–94.
30. Devede A, Pare D (1999) Impact of network activity on the integrative properties of neocortical pyramidal neurons in vivo. J Neurophysiol 81: 1531–1547.
31. Blaesse P, Guillemin I, Schindler J, Schweizer M, Delpire E, et al. (2006) Oligomerization of KCC2 correlates with development of inhibitory neurotransmission. J Neurosci 26: 10407–10419.
32. Uchavez P, Ludwig A, Marksten M, Sone S, Hubner CA, et al. (2009) Coexpression and heteromerization of two neuronal KCl cotransporter isoforms in neonatal brain. J Biol Chem 284: 13696–13704.
33. Godin AG, Costantino S, Lorenzo LE, Swift JL, Sergeev M, et al. (2011) Revealing protein oligomerization and densities in situ using spatial intensity distribution analysis. Proc Natl Acad Sci U S A 108: 7010–7015.
34. Swift JL, Godin AG, Dore K, Freland L, Bouchard N, et al. (2011) Quantification of receptor tyrosine kinase transactivation through direct dimmerization and surface density measurements in single cells. Proc Natl Acad Sci U S A 108: 7016–7021.
35. Costantino S, Krott CB, Godin AG, Kennedy TE, Wieman PW, et al. (2008) Semi-automated quantification of filopodial dynamics. J Neurosci Methods 171: 65–72.
36. Gulyas AI, Sik A, Payne JA, Kaila K, Freund TF (2001) The KCl cotransporter, KCC2, is highly expressed in the vicinity of excitatory synapses in the rat hippocampus. Eur J Neurosci 13: 2205–2217.
