

The Effect of Alterations of Schizophrenia-Associated Genes on Gamma Band Oscillations

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Abnormalities in the synchronized oscillatory activity of neurons in general and, specifically in the gamma band, might play a crucial role in the pathophysiology of schizophrenia. While these changes in oscillatory activity have traditionally been linked to alterations at the synaptic level, we demonstrate here, using computational modeling, that common genetic variants of ion channels can contribute strongly to this effect. Our model of primary auditory cortex highlights multiple schizophrenia-associated genetic variants that reduce gamma power in an auditory steady-state response task. Furthermore, we show that combinations of several of these schizophrenia-associated variants can produce similar effects as the more traditionally considered synaptic changes. Overall, our study provides a mechanistic link between schizophrenia-associated common genetic variants, as identified by genome-wide association studies, and one of the most robust neurophysiological endophenotypes of schizophrenia.

1 Introduction

The search for biological causes of psychiatric disorders has up to now met with limited success. While genetics and basic neuroscience have both made tremendous advances over the last decade, mechanistic links between genetic findings and clinical symptoms have so far not been discovered. Many have argued that symptom-based classifications of psychiatric illnesses might not be possible to map to alterations at the microscopic scale [18, 45], and have proposed to use biomarkers or endophenotypes, which in turn might correlate more clearly with genetic variants [45]. For example, the biomarkers and endophenotypes of schizophrenia (SCZ) include reduced mismatch negativity [100], reduced pre-pulse inhibition [96] and changes to evoked and induced oscillations

in multiple frequency bands in a large variety of tasks (e.g. [99]). Importantly, recent advances in computational modelling allow for the integration of knowledge about genetic contributions to ion channels and excitability and can be used to predict changes to macroscopic electroencephalography (EEG) or magnetoencephalography (MEG) signals ([35, 107, 60], for a review of this emerging subfield of computational psychiatry see [62]). For example, simulations of a detailed model of tufted layer 5 pyramidal cells have recently been used to predict the effect of SCZ-associated variants of ion channel-encoding genes on neural activity in the delta frequency band [63].

In general, oscillations in the low and high frequency range enable coordinated interactions between distributed neuronal responses [9, 27, 8, 87] and have been demonstrated to be functionally relevant [97]. For example, gamma oscillations have been linked to perception [34], attention [28], memory [92], consciousness [67] and synaptic plasticity [109]. In patients with schizophrenia, gamma power and coherence have been consistently found to be decreased during neural entrainment [51, 104, 49, 95] as well as during several sensory (e.g. visual Gestalt [90]) and cognitive (e.g. working memory [15]) tasks. Importantly, schizophrenia is also associated with disturbances in many of the above mentioned functions [77, 29, 98]. Mathematical analyses and computer simulations have demonstrated that gamma oscillations arise through the local interplay between excitatory and inhibitory populations, either through tonic excitation of inhibitory cells and subsequent rhythmic inhibition of excitatory cells (interneuron gamma or ING) or through rhythmic excitation of inhibitory cells and subsequent rhythmic inhibition of excitatory cells (pyramidal-interneuron gamma or PING) [111, 6]. The anatomical and electrophysiological properties of a particular subtype of inhibitory interneurons, the parvalbumin-positive (PV^+) interneurons, make them ideally suited for the fast, strong and temporally precise inhibition necessary for the generation of gamma rhythms [32]. Furthermore, optogenetically driving PV^+ interneurons was found to enhance gamma rhythms [11]. Consequently, cellular level alterations at PV^+ interneurons in schizophrenia have been linked to well-known auditory steady-state response (ASSR) deficits in the gamma band [104, 70, 71, 47, 86]. However, these studies have focused on changes to the strength and temporal dynamics of synaptic transmission. While synaptic transmission dynamics undoubtedly play a crucial role in the generation of neural oscillations, cell-intrinsic properties such as ionic conductances can also alter the ability of a network to generate and maintain oscillations. This is of particular importance since many of the recently discovered gene variants associated with schizophrenia relate to ionic channels or Ca^{2+} transporters [83, 22].

In this study, we use an established framework to translate the effect of common single-nucleotide polymorphism (SNP) variants associated with schizophrenia into a biophysically detailed model of a layer 5 pyramidal cell [60]. We then use a morphologically reduced version of the cell [59] together with a model of a PV^+ interneuron [103] in a microcircuit model and explore the effect of the genetic variants on gamma entrainment. We demonstrate that while single gene variants typically only have small effects on gamma auditory steady-state entrainment, combinations of them can reduce the entrainment comparable to the synaptic alterations mentioned above and replicate observations in schizophrenia patients. Our findings therefore provide a mechanistic link

between the scale of single genes and an important endophenotype of schizophrenia. Furthermore, the proposed model represents an ideal test-bed for the identification of targets for potential pharmacological agents aiming to reverse gamma deficits in schizophrenia.

2 Methods

2.1 Network Model

This study was based on a high-complexity, biophysically detailed model of thick-tufted layer 5 pyramidal cells [37]. The model includes a detailed reconstructed morphology, models of the dynamics of eleven different ionic channels and a description of the intracellular Ca^{2+} concentration [37]. Following earlier work, we incorporated human *in vitro* electrophysiological data on ion channel behavior from the functional genomics literature into this model [60, 63]. Due to the computational complexity of the original model, consisting of 196 compartments, we decided to use a reduced-morphology model, where passive parameters, ion channel conductances and parameters describing Ca^{2+} dynamics were fitted to reproduce the behaviour of the original model [59]. The inhibitory cells in the network were based on a model of fast-spiking PV^+ basket cells taken from [103]. These two single cell models were combined into a microcircuit network model consisting of 256 excitatory and 64 inhibitory neurons. Cells were connected via AMPA and NMDA receptor-mediated synaptic currents in the case of excitatory connections and GABA_A receptor-mediated synaptic currents for inhibitory connections. Additionally, model cells received two types of input, Poissonian noise to all cells representing background activity in the cortex and rhythmic input representing the sensory input during auditory entrainment. Note that a smaller percentage of inhibitory interneurons (35%) received no sensory input drive; this reflects preferential thalamic drive to pyramidal cell populations [4], which has also been used in other models [104]. This was to ensure that a subpopulation of the inhibitory neurons had a weak enough drive to be dominated by pyramidal cell activity. This subpopulation was necessary to maintain a 20 Hz peak for 40 Hz drive in of the synaptic alteration conditions against which we compared the genetic alterations. For a detailed discussion see the study by Vierling-Claassen et al. [104].

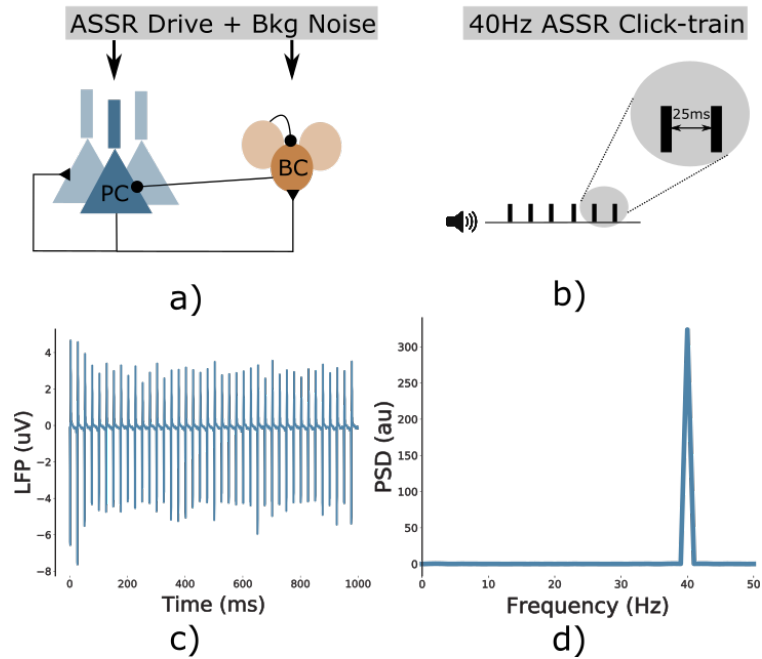


Figure 1: **Overview.** (a) Network schematic. The network consists of two interconnected populations, an excitatory population of pyramidal cells and an inhibitory population of basket cells, both receiving rhythmic ASSR drive and Poissonian background noise. (b) The 40 Hz ASSR drive is modeled as bouts of input spikes arriving at all cells simultaneously with an inter-bout interval of 25 ms, mimicking a 40 Hz ASSR click-train paradigm. (c) Example LFP signal of the control network in response to the ASSR drive, showing strong 40 Hz entrainment. (d) Power spectral density of the signal from c), again confirming that the network follows the 40 Hz click train rhythm.

2.2 Integration of Genetic Variants

We closely follow earlier work of [60, 64, 63] to integrate the effect of SNP-like genetic variants into our network model (details in Supplementary Section 4). In summary, we selected a set of genes, restricted to ion-channel-encoding genes likely to be expressed in layer 5 pyramidal cells, obtained from a large genome-wide association study (GWAS) [83]. For this set of genes, we searched the literature for genetic variants and their effects on electrophysiological parameters of pyramidal cells. This left us with 86 variants of the following genes: CACNA1C, CACNA1D, CACNB2, SCN1A and, HCN1 [50, 19, 40, 91, 94, 93, 5, 112, 79, 80, 3, 54, 17, 66, 55, 43, 14, 102, 105, 13, 65, 44, 53, 108]. These variants typically had a large effect on the electrophysiology of layer 5 pyramidal cells. However, due to the polygenic nature of SCZ, it can be assumed that the risk of the disorder is not caused by a single SCZ-associated SNP and, therefore, one would not expect large effects from the common variants identified in GWAS studies. Subsequently, we applied a downscaling procedure, as outlined in [60, 63]. In short, we downscaled the changes of model parameters induced by a variant (multiplication by a factor either on a linear or logarithmic scale, depending on the type of the parameter) until the cell response to predefined stimuli stayed within a certain range (Supplementary Section 4). This resulted in a set of 86 'small-effect' model variants, which were used as models for the effects of common variants on layer 5 pyramidal cell electrophysiological response features. As in earlier studies using this approach [60, 63], we will use the term 'variant' for a genetic variant in a human or animal genome and the term 'model variant' for a model of a gene variant constructed as described above.

3 Results

We executed simulations of the network model with background synaptic noise and a periodic drive at 40 Hz, mimicking auditory steady-state stimulation experiments. For each model variant in Supplementary table S1, we repeated the set of 200 simulations (10 'trials' for each of the 20 'virtual subjects', see Supplementary section 4). For each of these model variants, the parameters of the ion-channels were altered in a subtle way, leading to changes in their activation, and subsequently to modified network dynamics and gamma entrainment.

3.1 Single Variants Can Produce ASSR Power Deficits in the Gamma Range

First, we analyzed the evoked ASSR power in the gamma range. Most of the model variants altered gamma power in a weak or moderate way and the model variants could both increase or decrease gamma power (Figure 2). Model variants affecting the Na^+ channels had hardly any effect on entrainment power which reflects the small scaling coefficients imposed by the downscaling scheme (see Supplementary Table 3). Four model variants, one affecting the CACNA1C gene, one affecting the CACNA1I

gene and two affecting the HCN1A gene, led to strong decreases of evoked power (Figure 3). For the model variants affecting I_{CaHVA} , the change in evoked gamma power was positively correlated with the offset of the activation variable m (Pearson correlation: $r = 0.53$, $p < 0.001$; Supplementary Figure S9) but there was no significant correlation with other parameters describing the ion channel dynamics (Supplementary Figure S9). Furthermore, no model variant strongly increased the 20 Hz component (see Supplementary Figure S8), i.e. we observed no shift to the first subharmonic of the 40 Hz drive, which would be indicative of a 'beat-skipping' behaviour, where the inhibition suppresses every other drive stimulus as seen in models of altered synaptic dynamics [104].

3.2 Combinations of Variants Can Produce Substantial ASSR Power Deficits

Next, we tested three different combinations of model variants (details in Supplementary Tables 2 and 3). The combinations were: (1) the combination of the two model variants with the strongest gamma reduction from the single model variant trials (Ca7 affecting gene CACNA1C and HCN1-2 affecting HCN1; this combination will be referred to as *Comb1*); (2) the model variants from (1) but additionally one more model variant with a moderate gamma reduction (Ca74 affecting CACNA1I gene, this combination will be referred to as *Comb2*); (3) the model variants from Comb2 plus an additional model variant with a moderate gamma reduction (HCN1-1 affecting the HCN1 gene; this combination will be referred to as *Comb3*). In the last case, when combining model variants of the same gene, we assume a linear superposition of the effects of the single model variants on the parameters of the ion channels. Note that this is a simplistic assumption and that there could potentially be nonlinear interactions between different variants of the same gene. However, actual experimental data on this relationship are currently not available. We found that combining model variants further increased their effect on evoked gamma power and a combination of only a few variants already had a strong impact on gamma power (Figure 4 and Supplementary Table S4). Overall, the effects of combining model variants were additive, e.g. combining the two model variants Ca7 and HCN1-2 which individually have a mean difference of -29.5 and -21.9 results in a combination with a mean difference of -49.7 which is roughly the sum of the two individual mean differences (see also Supplementary Table 4).

3.3 The Effects of Variant Combinations Are Comparable to the Effects of Synaptic Alterations

Changes to the GABAergic system have been proposed to explain the reduction in evoked gamma power [32] and modelling work has demonstrated that both, a reduction in GABA levels in schizophrenia patients [89, 70] and an increase in decay times at GABAergic synapses [104, 70], can lead to reduced ASSR gamma power. On the one hand, a decrease in inhibition due to reduced GABA levels decreases the precise inhibitory control over pyramidal cell firing necessary for a strong gamma rhythm. On

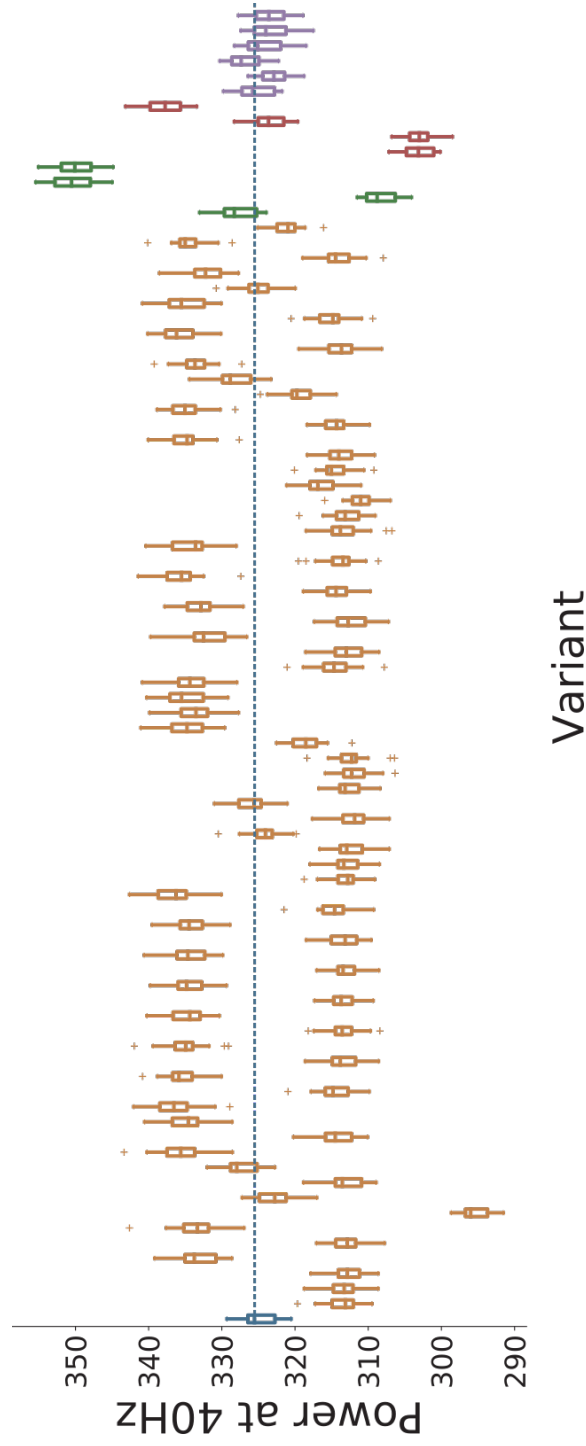


Figure 2: **Overview of the 4040 measure for all variants.** Control network in blue, Ca^{2+} channel variants affecting Ca_{HVA} in orange, Ca^{2+} channel variants affecting Ca_{LVA} in green, HCN variants in red, and SCN variants in purple. Solid lines represent the mean, box edges the 25 and 75 percentile, respectively, the whisker extend to 2 standard deviations and + depict outliers. The dashed blue line represents the mean of the control network.

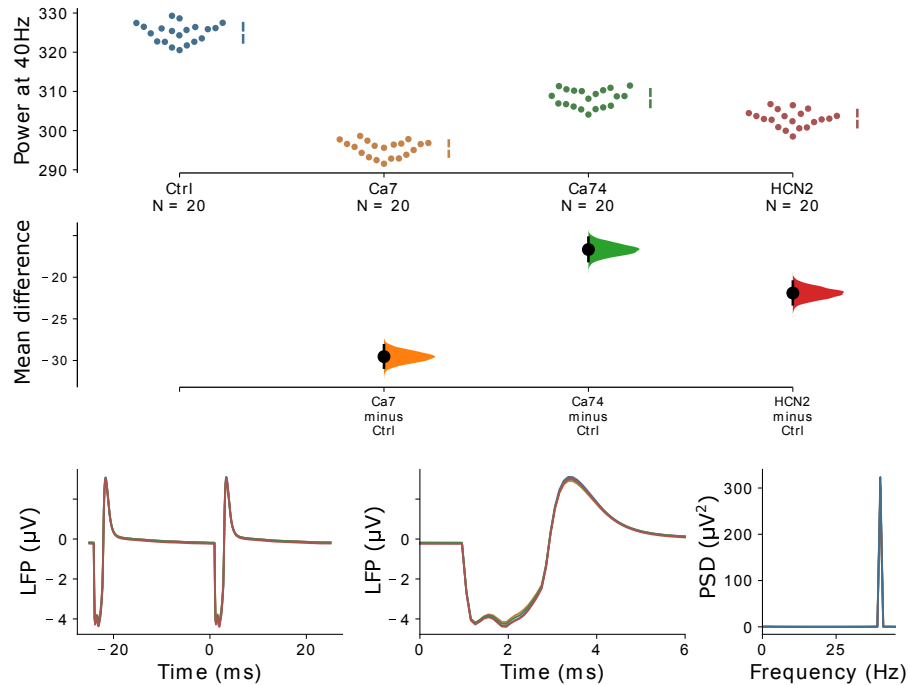


Figure 3: **Comparison of different variants against the control.** (a) The mean difference, i.e. for three comparisons (the Ca^{2+} channel model variants, Ca7 and Ca74, and the HCN model variant, HCN1-2, with the strongest gamma reduction) against the shared control are shown in the above Cumming estimation plot. The raw data is plotted on the upper axes. On the lower axes, mean differences are plotted as bootstrap sampling distributions. Each mean difference is depicted as a dot. Each 95% confidence interval is indicated by the ends of the vertical error bars. (b) The simulated LFP signal for the control network (blue), Ca7 (yellow), Ca74 (green), and HCN1-2 (red) is shown, averaged over two consecutive gamma cycles. (c) The two signals from (b) are presented with a zoom into the narrow time frame after the stimulus arrives at 0 ms. (d) The power spectral density (PSD) for the LFP signals from (b). Note that in (b)-(d) the LFP signal is first averaged over all 'subjects' and 'trials'.

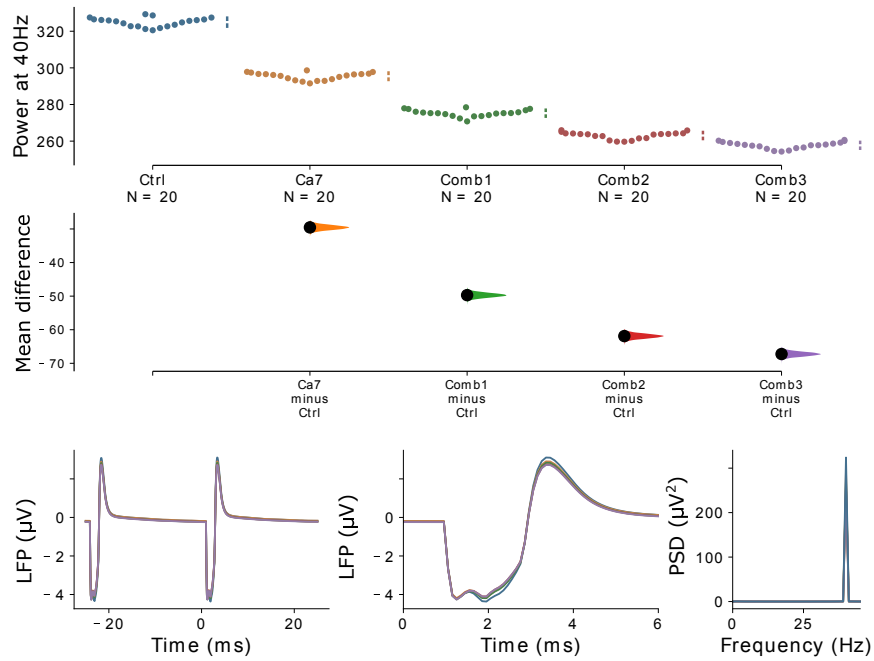


Figure 4: **Comparison of different combinations of variants against the control.** The mean difference for the comparisons of three variant combinations and the single variant with the strongest effect (Ca7) against the shared control are shown in the above Cumming estimation plot. The raw data is plotted on the upper axes. On the lower axes, mean differences are plotted as bootstrap sampling distributions. Each mean difference is depicted as a dot. Each 95% confidence interval is indicated by the ends of the vertical error bars.

the other hand, an increase in GABAergic decay times, while increasing inhibition, can lead to suppression of pyramidal cell firing every second gamma cycle (a phenomenon called 'beat-skipping') and thus strongly reduce gamma power. Therefore, we compared the model variant effects to the effects of these two changes to the GABAergic system of the network. Specifically, we implemented a 25% reduction of the maximum conductance of GABAergic synapses (see e.g. [70]) and an increase of the decay time of inhibitory postsynaptic currents (IPSCs) at GABAergic synapses from 8 ms to 25 ms (e.g. [104, 70]). These two conditions will be referred to as *Gmax* and *IPSC*. Comparing the effects of the model variants and their combinations to these two changes allowed us to judge the relative size of their effect.

Figure 5 shows that, while even the strongest individual model variants only result in moderate reductions of gamma power, combinations of several model variants can produce strong gamma reduction comparable to a strong reduction of GABA levels (see also Supplementary Table 4). Interestingly, the increase in GABAergic decay times produced a substantially stronger reduction of gamma power, mainly due to the emergence of a beat-skipping behaviour, which shifted the power from the 40 Hz gamma band to the subharmonic 20 Hz beta band (see Supplementary Figure 8). Nevertheless, these results demonstrate that combinations of model variants, influencing ionic channels and single cell excitability, can potentially have strong effects on gamma entrainment.

3.4 Genetic Variants Do not Affect the Inter-Trial Phase Coherence

Besides strong reductions of evoked power, many studies report a decrease in inter-trial phase coherence (ITPC), a measure of how aligned the phase angles of the signal over individual trials are, during gamma entrainment in patients with schizophrenia (e.g. [51, 49], see Thune et al. [95] for a review). Therefore, we also calculated the ITPC for each single model variant and compared them to the two synaptic conditions from before (see 2). We found that none of the model variants altered the phase coherence substantially, while both synaptic conditions strongly decreased ITPC (Figure 6). This means that while both synaptic conditions reduce phase coherence, i.e. desynchronize the network, the reduction in gamma power resulting from the model variants seems to come solely from a reduction in amplitude of the entrained oscillations, leaving the temporal precision intact. Furthermore, this suggests that, while genetic variants of ion channel-encoding genes might contribute to the reductions in evoked gamma power found in patients with schizophrenia, it is unlikely that they play an important role in the emergence of decreases in phase coherence and the underlying desynchronization of activity in the network.

3.5 Correlations to Pre-Pulse Inhibition and Delta Resonance

Lastly, we compared the effect of the genetic variants with their effect on other potential biomarkers for SCZ from our earlier studies [63]. We calculated Pearson correlation coefficients between the ratio of gamma reduction and the delta resonance power and

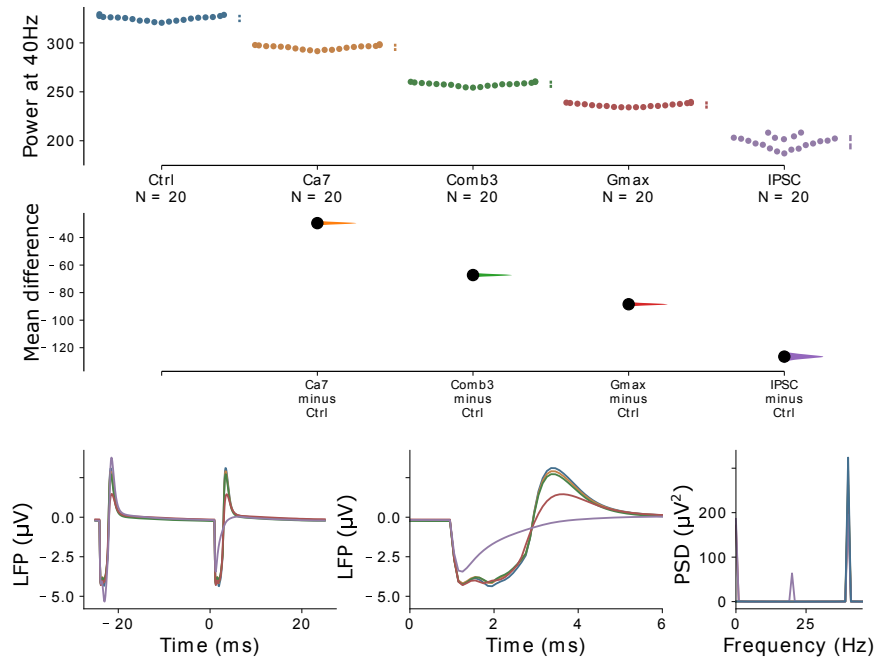


Figure 5: **Comparison of model variants against synaptic alterations.** The mean difference for the comparisons of the single model variant and the model variant combination with the strongest gamma reduction together with the two synaptic mechanisms, *Gmax* and *IPSC*, against the shared control are shown in the above Cumming estimation plot. The raw data is plotted on the upper axes. On the lower axes, mean differences are plotted as bootstrap sampling distributions. Each mean difference is depicted as a dot. Each 95% confidence interval is indicated by the ends of the vertical error bars.

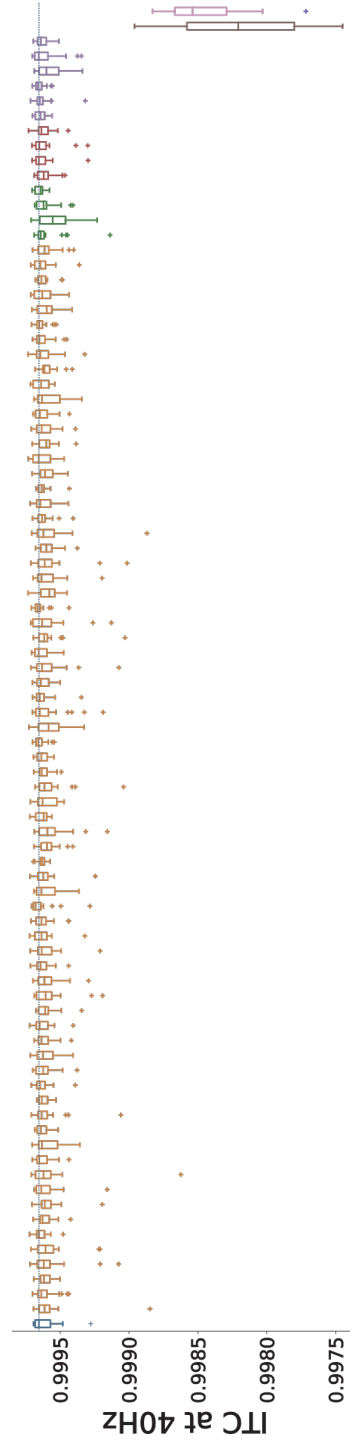


Figure 6: **Overview of 4040 inter-trial coherence for all variants.** Control network in blue, Ca^{2+} channel model variants affecting Ca_{HVA} in yellow, Ca^{2+} channel model variants affecting Ca_{LVA} in green, HCN model variants in red, SCN model variants in purple, the reduction of g_{max} in brown and increased IPSC times in steel pink. Solid line represents the mean, box edges the 25 and 75 percentile, respectively, the whisker extend to 2 standard deviations and + depict outliers. the dashed blue line depicts the mean of the control network.

the pre-pulse inhibition (PPI) thresholds for all 86 model variants (see Supplementary Methods for details). We found a strong positive correlation between gamma power and delta resonance power (Pearson $r=0.683$, $p<0.0001$) and moderate negative correlation between gamma power and PPI threshold (Pearson $r=-0.365$, $p<0.0001$).

4 Discussion

In this modelling study we showed that changes to the kinetics of voltage-gated ion channels due to SCZ-associated common variants of their encoding genes led to decreases of evoked gamma power, one of the most frequently reported electrophysiological phenotypes in SCZ (Figures 2 and 3). We further demonstrated that combinations of model variants produced larger decreases in evoked gamma power (Figure 4) and that these decreases were comparable to alterations at the synaptic level (Figure 5), which are more commonly associated with changes of evoked oscillatory power in the gamma range [104, 70]. Interestingly, we found that the genetic variants, opposed to the synaptic alterations, did not change the inter-trial phase coherence, a measure of synchronization. The reductions in evoked gamma power were solely due to reductions in the amplitude of the oscillations (Figure 5).

The genetic variants that decreased evoked gamma power identified in this study mostly decreased the excitability of the layer 5 pyramidal cell, as found in earlier work [60], and, therefore, led to smaller amplitudes of the evoked LFP signals. Not surprisingly, there was not much overlap with the variants we previously found to substantially increase network delta oscillations and to reduce single cell PPI [63]. We further analysed this by correlating gamma power with delta resonance power and PPI threshold from this previous study, respectively. Here, we found a strong positive correlation between gamma and delta resonance power and a moderate negative correlation with PPI thresholds. This suggests that variants decreasing gamma power also lead to lower delta resonance power and larger PPI thresholds. While robust evidence for decreased PPI thresholds in SCZ exists [96], the findings on delta power show mixed results. Several studies find increased delta power in patients [72, 7, 41, 36, 24], however, decreased delta oscillation power has also been reported [23, 26]. Nonetheless, as also described earlier, it seems unlikely that gamma power reduction, delta power changes and PPI threshold decrease are solely caused by genetic variants affecting ionic channels. Gamma reductions most certainly are at least partially attributable to synaptic alterations as explained earlier. Nevertheless, the variants modelled here might play an important role in gamma reduction in subpopulations of patients and contribute to the large heterogeneity observed in patients with schizophrenia.

Previous models of gamma range oscillatory deficits in SCZ have mainly focused on changes at the synaptic level, such as changes of GABAergic synapses from PV⁺ interneurons onto pyramidal cells or other PV⁺ interneurons [70, 71, 104, 86, 89, 106], changes of glutamatergic excitation of PV⁺ interneurons through NMDA receptors, [46, 86] or changes of spine density at pyramidal cells [86]. As mentioned in the Results section, there are two main consequences of changes to the GABAergic system that have been the focus of previous modelling studies: 1) A reduction of the peak

amplitude of the IPSC and 2) a prolongation of IPSC decay times. A reduction of the peak IPSC amplitude has been shown to significantly reduce evoked gamma power but to leave beta power intact [70, 86, 89]. On the other hand, an increase of IPSC decay time, while also substantially reducing evoked gamma power, has been shown to increase power in the beta band [104, 70] and most probably exerts its effect through PV⁺ basket cells [71]. Kirli et al. [46, 48] demonstrated that lower gamma band power was present at both low and high NMDA conductance levels with optimal synchronization occurring at intermediate conductance levels. In another study, Siekmeier and van Maanen [86] showed that modest reductions in NMDA system function and dendritic spine density led to a robust reduction of gamma power. However, they also found that greater NMDA hypofunction along with low level GABA system dysregulation substantially decreased gamma power, highlighting the multifactoriality of underlying alterations. In addition, dopaminergic modulation of local circuits has also been shown to affect gamma synchronization through modulation of K⁺ currents and NMDA conductances [47]. Overall we must note that, as demonstrated for models of gamma ASSRs in particular [86, 70] but also other models of psychiatric disorders [78] and models of healthy local circuits [81] in general, many different parameter combinations might produce similar network level behaviour. Therefore, it is very important to explore the interaction of alterations of different systems. Moreover, to constrain the models further, these interactions should be tested against different paradigms, such as spontaneous and evoked oscillations. The computational model presented here offers an ideal starting point for such an effort, since it allows for the integration of the most crucial factors contributing to gamma band oscillatory deficits in schizophrenia. Beyond incorporating variants of ion channel encoding genes and alterations of GABAergic synapses, extensions of the model could include the integration of NMDAR hypofunction via its NMDARs and the integration of changes to dopaminergic neuromodulation via its K⁺ channels, for example as in [47].

A crucial part of the overall approach used in this study is the dampening of the effect of literature-derived model variants by a downscaling of the parameter changes. Typically, the literature-derived model variants very strongly changed the physiology of the studied cell (see also [60]) and, therefore, were not representative of common SNP-like variants. Overall, single SNP-like variants, which are known to be numerous and to occur frequently in the healthy population [83], are assumed to have small phenotypic effects, either on the single cell or on the systems level. Nevertheless, we note that rare variants with large effects associated with schizophrenia also exist (see e.g. [2]). Additionally, we have shown in previous work on increased delta oscillations due to genetic variants that model variants derived from gene expression data largely result in similar changes to the network model [63].

GWAS studies, such as Ripke et al. [83], have identified numerous variants associated with psychiatric disorders, however, we know very little about their functional effects. As we have argued before [63, 62], the modelling framework presented in this study is ideally suited to build hypotheses about their effects and to make experimentally testable predictions. To be more specific, the biophysically detailed model used here can provide very specific associations between genetic variants and phenotypes, while explicitly revealing the cellular properties through which the two are mechanistically

linked. This goes well beyond the purely statistical associations that standard genetics approaches produce.

Limitations and future directions The analysis presented here is based on the model of a thick-tufted layer 5 pyramidal cell, which accurately reproduces many active and passive electrophysiological features of these cells [37], and the effects of the genetic variants were implemented as changes to the kinetics of the underlying ion channels of the model. Therefore, our approach here rests on the assumption that the model faithfully reproduces the ion channel dynamics of layer 5 pyramidal cells. While the fitting of the model did not include the replication of activity in the presence of ion-channel blockers [58], the model's ion channel composition is largely consistent with that of other models. Almog et al. present a model of a layer 5 pyramidal cell where the set of channels is partly overlapping [1]. Although the contributions of the ion channels to model behaviour differ slightly, with the persistent K^+ having a larger role, the two models mostly conform with each other [1, 64]. Papoutsi et al. [76] also present a model of layer 5 pyramidal cells showing similar interactions between voltage-gated Ca^{2+} channels and the Ca^{2+} -dependent afterhyperpolarization (AHP) current as in the model of Hay et al. we used here and further underpin the validity of our model assumptions.

While in this study we have focused on the effect of genetic variants on ion channel dynamics, future studies should also explore their role in schizophrenia-associated changes to synaptic receptors, especially GABA and NMDA receptors. These are, as mentioned before, more traditionally associated with gamma band deficits in the disorder [32, 33]. Furthermore, the approach outlined in this study could also be extended to study the effect of genetic variants on intracellular signaling cascades involved in plasticity [22, 61]. A major challenge for the field, however, will be to incorporate immune pathways, which have been strongly indicated by recent GWAS [83, 101], into models of schizophrenia pathophysiology.

Nevertheless, the model developed in this study can already be used to complement more traditional approaches to identify potential treatment targets. Current medications with known courses of action can be included into the model and their effects on gamma band oscillations can subsequently be assessed (see [86] for an excellent implementation of such an approach). Furthermore, a search for novel therapeutic targets can be conducted through a more exploratory analysis of the effect of parameters on the model behaviour (also see [86]). Additionally, such an approach is not limited to pharmacological interventions since transcranial electric or magnetic stimulation can easily be incorporated (as demonstrated in other modeling studies such as [10, 85, 82]); this is, of course, not restricted to schizophrenia but can be applied to psychiatric disorders in general.

In conclusion, our work represents a step towards the integration of the wealth of genetic data on psychiatric disorders into biophysically detailed models of biomarkers with great potential to unravel underlying polygenic cellular-based mechanisms. Furthermore, the approach offers an ideal test ground for the identification of novel therapeutic strategies, such as pharmaceutical interventions or electrical stimulation.

Supplementary Material

Supplementary Methods

Ion channels and their genetic etiology

The single cell model used to model the pyramidal cells in our network is based on a detailed, multi-compartment Hodgkin-Huxley type model of layer 5 pyramidal cells with a reconstructed morphology [37]. However, the very high morphological detail (196 compartments), and thus very high computational complexity, of this neuron model renders it unsuitable for the network model analysis performed in this study. Therefore, we employed a reduced version of this model, where passive parameters and ion channel and Ca^{2+} dynamics were fitted to the original model using a multi-step fitting procedure [59]. The reduced model, analogous to the original model, contains the following ionic currents: Fast inactivating Na^+ current (I_{Nat}), persistent Na^+ current (I_{Nap}), non-specific cation current (I_h), muscarinic K^+ current (I_m), slow inactivating K^+ current (I_{Kp}), fast inactivating K^+ current (I_{Kt}), fast non-inactivating K^+ current ($I_{\text{Kv3.1}}$), high-voltage-activated Ca^{2+} current (I_{CaHVA}), low-voltage-activated Ca^{2+} current (I_{CaLVA}), small-conductance Ca^{2+} -activated K^+ current (I_{SK}), and the passive leak current (I_{leak}).

Exact information on which ion channel subunits contribute to the above mentioned currents in layer 5 pyramidal cells is still missing. However, the expression of different ion channel subunits in these cells has been studied extensively, and indications can be drawn from these studies. Christophe et al. [16] found the mRNA of the ion channel-encoding genes *KCNA2*, *KCND2*, *KCND3*, *CACNA1A*, *CACNA1B*, *CACNA1C*, *CACNA1D*, *CACNA1E*, *CACNA1G*, *CACNA1H*, *CACNA1I*, *HCN1*, and *HCN2* expressed in postnatal rat neocortices at different stages of development. Of these genes, it is known that *CACNA1A*, *CACNA1B*, *CACNA1C*, and *CACNA1D* contribute to I_{CaHVA} , *CACNA1G*, *CACNA1H*, and *CACNA1I* to I_{CaLVA} , while the genes *KCNA2*, *KCND2*, and *KCND3* might contribute to the slow I_{Kp} current.

In a different study, the genes *SCN1A*, *SCN2A*, *SCN3A*, and *SCN6A* were found to be expressed in layer 5 pyramidal cells [110]. While several of the genes encoding the α subunits (*SCN1A*, *SCN2A* and *SCN3A*) are tetrodotoxin-sensitive [12] and hence form both the transient (I_{Nat}) and persistent (I_{Nap}) Na^+ currents, their contribution to I_{Nat} or I_{Nap} might be dependent on modulatory subunits [56].

Description of the Single Cell Models

Layer 5 pyramidal cells The layer 5 pyramidal cell model is a multi-compartment Hodgkin-Huxley type model and the membrane potential dynamics can be described by

$$C_m \frac{\partial V}{\partial t} = I_{\text{Nat}} + I_{\text{Nap}} + I_h + I_m + I_{\text{Kp}} + I_{\text{Kt}} + I_{\text{Kv3.1}} + I_{\text{CaHVA}} + I_{\text{CaLVA}} + I_{\text{SK}} + I_l + I_{\text{axial}},$$

where the different currents can be written as the product of an activation and an inactivation variable

$$I = \bar{g}m^{N_m}h^{N_h}(E - V).$$

Here, m and h are the activation and inactivation variables, N_m and N_h their sensitivities, \bar{g} the maximal ionic conductance and E the ionic reversal potential. Na^+ and K^+ have a fixed reversal potential of $E_{\text{Na}} = 50 \text{ mV}$ and $E_{\text{K}} = -85 \text{ mV}$, respectively, while the reversal potential of Ca^{2+} depends on the intracellular $[\text{Ca}^{2+}]$. Furthermore, the dynamics of activation and inactivation are defined by

$$\frac{dm}{dt} = \frac{m - m_\infty}{\tau_m} \quad \text{and} \quad \frac{dh}{dt} = \frac{h - h_\infty}{\tau_h},$$

where m_∞ , h_∞ , τ_m , and τ_h are functions of the membrane potential V . In our case, m_∞ and h_∞ usually have a sigmoidal shape, where the half-activation and half-inactivation voltages are determined by one or more parameters each (depending on the ion channel). These parameters are written as $V_{\text{offm}*}$ and $V_{\text{offh}*}$, where $*$ stands for further specifications in case there are multiple parameters affecting them. Analogously, the parameters $V_{\text{slo}*}$ and $V_{\text{sloh}*}$ affect the slopes of the (in)activation curves, and parameters $\tau_{\text{m}*}$ and $\tau_{\text{offh}*}$ the time constants. All ionic currents, except for the activation of I_{SK} , are described in this way. The activation of I_{SK} , on the other hand, only depends on the intracellular $[\text{Ca}^{2+}]$, through a sigmoidal function with a half-activation concentration parameter c_{off} and a slope parameter c_{slo} . Lastly, the dynamics of the intracellular $[\text{Ca}^{2+}]$ is described by

$$\frac{d[\text{Ca}^{2+}]_i}{dt} = \frac{I_{\text{CaHVA}} + I_{\text{CaLVA}}}{2\gamma Fd} - \frac{[\text{Ca}^{2+}]_i - c_{\text{min}}}{\tau_{\text{decay}}}, \quad (1)$$

where I_{CaHVA} and I_{CaLVA} are the high and low-voltage activated Ca^{2+} currents entering the considered cell segment, γ represents the fraction of Ca^{2+} ions entering the cell that contribute to the intracellular $[\text{Ca}^{2+}]$, F the Faraday constant, d is the depth of the sub-membrane layer considered for calculation of concentration, c_{min} the resting intracellular $[\text{Ca}^{2+}]$, and τ_{decay} is the decay time constant of the intracellular $[\text{Ca}^{2+}]$.

As mentioned earlier, in this study we employed the reduced model presented in [59], consisting of four compartments: the soma, the apical trunk, the apical tuft, and the basal dendrite. the maximal conductance values \bar{g} were different for different compartments and are subindexed as follows: s refers to the soma, a_0 to the apical trunk, a_1 to the apical tuft, and b to the basal dendrite. All values not shown are set to 0.

Fast inactivating Na^+ current, I_{Nat}

$$\alpha_m = -\frac{1}{\tau_{\text{ma}}} \cdot \frac{V_{\text{offm}} - V}{1 - \exp\left(\frac{V_{\text{offm}} - V}{V_{\text{slo}*}}\right)}$$

$$\beta_m = \frac{1}{\tau_{\text{mb}}} \cdot \frac{V_{\text{offm}} - V}{1 - \exp\left(\frac{-(V_{\text{offm}} - V)}{V_{\text{slo}*}}\right)}$$

$$\alpha_h = \frac{1}{\tau_{ha}} \cdot \frac{V_{offh} - V}{1 - \exp\left(\frac{(V_{offh}-V)}{V_{sloh}}\right)}$$

$$\beta_h = -\frac{1}{\tau_{hb}} \cdot \frac{V_{offh} - V}{1 - \exp\left(\frac{-(V_{offh}-V)}{V_{sloh}}\right)}$$

$$m_\infty = \frac{\alpha_m}{\alpha_m + \beta_m}$$

$$h_\infty = \frac{\alpha_h}{\alpha_h + \beta_h}$$

$$\tau_m = \frac{1}{T_{adj}(\alpha_m + \beta_m)}$$

$$\tau_h = \frac{1}{T_{adj}(\alpha_h + \beta_h)}$$

$V_{offm} = -38$ mV, $V_{offh} = -66$ mV, $V_{sлом} = 6.0$ mV, $V_{sloh} = 6.0$ mV, $\tau_{ma} = 5.49$ ms, $\tau_{mb} = 8.06$ ms, $\tau_{ha} = 66.67$ ms, $\tau_{hb} = 66.67$ ms, $\bar{g}_s = 2.41$ S/cm², $\bar{g}_{a_0} = 0.0135$ S/cm², $\bar{g}_{a_1} = 0.0131$ S/cm², $N_m = 3$, $N_h = 1$

Persistent Na⁺ current, I_{Nap}

$$m_\infty = \frac{1}{1 + \exp\left(\frac{V_{offm}-V}{V_{sлом}}\right)}$$

$$h_\infty = \frac{1}{1 + \exp\left(-\frac{V_{offh}-V}{V_{sloh}}\right)}$$

$$\alpha_m = -\frac{1}{\tau_{ma}} \cdot \frac{V_{offma} - V}{1 - \exp\left(\frac{(V_{offma}-V)}{V_{sломa}}\right)}$$

$$\beta_m = \frac{1}{\tau_{mb}} \cdot \frac{V_{offmb} - V}{1 - \exp\left(\frac{-(V_{offmb}-V)}{V_{sломb}}\right)}$$

$$\alpha_h = \frac{1}{\tau_{ha}} \cdot \frac{V_{offha} - V}{1 - \exp\left(-\frac{V_{offha}-V}{V_{slohа}}\right)}$$

$$\beta_h = -\frac{1}{\tau_{hb}} \cdot \frac{V_{offhb} - V}{1 - \exp\left(\frac{V_{offhb}-V}{V_{slohь}}\right)}$$

$$\tau_m = \frac{6}{T_{adj}(\alpha_m + \beta_m)}$$

$$\tau_h = \frac{1}{T_{adj}(\alpha_h + \beta_h)}$$

$V_{offm} = -52.6$ mV, $V_{sлом} = 4.6$ mV, $V_{offma} = -38$ mV, $V_{offmb} = -38$ mV, $V_{sломa} = 6.0$ mV, $V_{sломb} = 6.0$ mV, $\tau_{ma} = 5.49$ ms, $\tau_{mb} = 8.06$ ms, $V_{offh} = -48.8$ mV, $V_{sloh} = 10.0$ mV, $V_{offha} = -17$ mV, $V_{offhb} = -64.4$ mV, $V_{slohа} = 4.63$ mV, $V_{slohь} = 2.63$ mV,

$$\tau_{ha} = 347222.2 \text{ ms}, \tau_{hb} = 144092.2 \text{ ms}, \bar{g}_s = 0.00206 \text{ S/cm}^2, N_m = 3, N_h = 1$$

Non-specific cation current, I_h

$$\alpha_h = -\frac{1}{\tau_{ha}} \cdot \frac{V_{\text{offha}} - V}{\exp\left(-\frac{V_{\text{offha}} - V}{V_{\text{sloha}}}\right) - 1}$$

$$\beta_h = \frac{1}{\tau_{hb}} \exp\left(-\frac{V_{\text{offhb}} - V}{V_{\text{slohb}}}\right)$$

$$h_\infty = \frac{\alpha_h}{\alpha_h + \beta_h}$$

$$\tau_h = \frac{1}{\alpha_h + \beta_h}$$

$$E = -45.0 \text{ mV}, V_{\text{offha}} = -154.9 \text{ mV}, V_{\text{sloha}} = 11.9 \text{ mV}, \tau_{ha} = 155.52 \text{ ms}, V_{\text{offhb}} = 0.0 \text{ mV}, V_{\text{slohb}} = 33.1 \text{ mV}, \tau_{hb} = 5.18 \text{ ms}, \bar{g}_s = 0.000279 \text{ S/cm}^2, \bar{g}_{a_1} = 0.00493 \text{ S/cm}^2, \bar{g}_b = 0.000294 \text{ S/cm}^2, N_m = 0, N_1 = 0$$

Muscarinic K^+ current, I_m

$$\alpha_m = \frac{1}{\tau_{ma}} \exp\left(-\frac{V_{\text{offma}} - V}{V_{\text{sloa}}}\right)$$

$$\beta_m = \frac{1}{\tau_{mb}} \exp\left(\frac{V_{\text{offmb}} - V}{V_{\text{slob}}}\right)$$

$$m_\infty = \frac{\alpha_m}{\alpha_m + \beta_m}$$

$$\tau_m = \frac{1}{T_{\text{adj}}(\alpha_m + \beta_m)}$$

$$V_{\text{offma}} = -35 \text{ mV}, V_{\text{sloa}} = 10 \text{ mV}, \tau_{ma} = 303.03 \text{ ms}, V_{\text{offmb}} = -35 \text{ mV}, V_{\text{slob}} = 10 \text{ mV}, \tau_{mb} = 303.03 \text{ ms}, \bar{g}_{a_0} = 0.000143 \text{ S/cm}^2, \bar{g}_{a_1} = 0.000113 \text{ S/cm}^2, N_m = 1, N_h = 0$$

Slow inactivating K^+ current, I_{Kp}

$$m_\infty = \frac{1}{1 + \exp\left(\frac{V_{\text{offm}} - V}{V_{\text{slo}}}\right)}$$

$$h_\infty = \frac{1}{1 + \exp\left(-\frac{V_{\text{offh}} - V}{V_{\text{sloh}}}\right)}$$

$$\tau_m = \begin{cases} \frac{\tau_{\text{mmin}} + \tau_{\text{mdiff1}} \exp\left(-\frac{V_{\text{offmt}} - V}{V_{\text{slomt}}}\right)}{T_{\text{adj}}}, & \text{if } V \leq V_{\text{thresh}} \\ \frac{\tau_{\text{mmin}} + \tau_{\text{mdiff2}} \exp\left(\frac{V_{\text{offmt}} - V}{V_{\text{slomt}}}\right)}{T_{\text{adj}}}, & \text{if } V > V_{\text{thresh}} \end{cases}$$

$$\tau_h = \frac{\tau_{\text{hmin}} + (\tau_{\text{hdiff1}} - \tau_{\text{hdiff2}}(V_{\text{offht1}} - V)) \exp\left(-\left(\frac{V_{\text{offht2}} - V}{V_{\text{sloht}}}\right)^2\right)}{T_{\text{adj}}}$$

$$V_{thresh} = V_{offmt} - \frac{V_{sloimt}}{2} \log \left(\frac{\tau_{mdiff1}}{\tau_{mdiff2}} \right)$$

$V_{offm} = -11$ mV, $V_{sloim} = 12$ mV, $V_{offmt} = -10$ mV, $V_{sloimt} = 38.46$ mV, $\tau_{mmin} = 1.25$ ms, $\tau_{mdiff1} = 175.03$ ms, $\tau_{mdiff2} = 13$ ms, $V_{offh} = -64$ mV, $V_{sloih} = 11$ mV, $V_{offht1} = -65$ mV, $V_{offht2} = -85$ mV, $V_{sloiht} = 48$ mV, $\tau_{hmin} = 360$ ms, $\tau_{hdiff1} = 1010$ ms, $\tau_{hdiff2} = 24$ ms/mV, $\bar{g}_s = 0.000176$ S/cm², $N_m = 2$, $N_h = 1$

Fast inactivating K⁺ current, I_{Kt}

$$m_{\infty} = \frac{1}{1 + \exp\left(\frac{V_{offm} - V}{V_{sloim}}\right)}$$

$$h_{\infty} = \frac{1}{1 + \exp\left(-\frac{V_{offh} - V}{V_{sloih}}\right)}$$

$$\tau_m = \frac{\tau_{mmin} + \tau_{mdiff} \exp\left(-\left(\frac{V_{offmt} - V}{V_{sloimt}}\right)^2\right)}{T_{adj}}$$

$$\tau_h = \frac{\tau_{hmin} + \tau_{hdiff} \exp\left(-\left(\frac{V_{offht} - V}{V_{sloiht}}\right)^2\right)}{T_{adj}}$$

$V_{offm} = -10$ mV, $V_{sloim} = 19$ mV, $V_{offh} = -76$ mV, $V_{sloih} = 10$ mV, $V_{offmt} = -81$ mV, $V_{sloimt} = 59$ mV, $\tau_{mmin} = 0.34$ ms, $\tau_{mdiff} = 0.92$ ms, $V_{offht} = -83$ mV, $V_{sloiht} = 23$ mV, $\tau_{hmin} = 8$ ms, $\tau_{hdiff} = 49$ ms, $\bar{g}_s = 0.0239$ S/cm², $N_m = 4$, $N_h = 1$

Fast, non inactivating K⁺ current, $I_{Kv3.1}$

$$m_{\infty} = \frac{1}{1 + \exp\left(\frac{V_{offm} - V}{V_{sloim}}\right)}$$

$$h_{\infty} = \frac{1}{T_{adj} \left(1 + \exp\left(\frac{V_{offh} - V}{V_{sloih}}\right)\right)}$$

$V_{offma} = 18.7$ mV, $V_{offmt} = -46.56$ mV, $V_{sloima} = 9.7$ mV, $V_{sloimt} = 44.14$ mV, $\tau_{mmax} = 4.0$ ms, $\bar{g}_s = 0.701$ S/cm², $\bar{g}_{a_0} = 0.00121$ S/cm², $N_m = 1$, $N_h = 0$

High-voltage-activated Ca²⁺ current, I_{CaHVA}

$$\alpha_m = -\frac{1}{\tau_{ma}} \cdot \frac{V_{offma} - V}{1 - \exp\left(\frac{V_{offma} - V}{V_{sloima}}\right)}$$

$$\beta_m = \frac{1}{\tau_{mb}} \exp\left(-\frac{V_{offmb} - V}{V_{sloimb}}\right)$$

$$m_{\infty} = \frac{\alpha_m}{\alpha_m + \beta_m}$$

$$\begin{aligned}\tau_m &= \frac{1}{\alpha_m + \beta_m} \\ \alpha_h &= \frac{1}{\tau_{ha}} \exp\left(\frac{V_{offha} - V}{V_{sloha}}\right) \\ \beta_h &= -\frac{1}{\tau_{hb}} \cdot \frac{1}{1 + \exp\left(\frac{(V_{offhb} - V)}{V_{slohb}}\right)} \\ h_\infty &= \frac{\alpha_h}{\alpha_h + \beta_h} \\ \tau_h &= \frac{1}{\alpha_h + \beta_h}\end{aligned}$$

$V_{offma} = -27$ mV, $V_{offmb} = -75$ mV, $V_{offha} = -13$ mV, $V_{offhb} = -15$ mV, $V_{sloa} = 3.8$ mV, $V_{slohb} = 17$ mV, $V_{sloha} = 50$ mV, $V_{slohb} = 28$ mV, $\tau_{ma} = 18.18$ ms, $\tau_{mb} = 1.06$ ms, $\tau_{ha} = 2188.18$ ms, $\tau_{hb} = 153.85$ ms, $\bar{g}_s = 0.000838$ S/cm², $\bar{g}_{a_1} = 0.000977$ S/cm², $N_m = 2$, $N_h = 1$

Low-voltage-activated Ca²⁺ current, I_{CaLVA}

$$\begin{aligned}m_\infty &= \frac{1}{1 + \exp\left(\frac{V_{offm} - V}{V_{slo}}\right)} \\ h_\infty &= \frac{1}{1 + \exp\left(-\frac{V_{offh} - V}{V_{sloh}}\right)} \\ \tau_m &= \tau_{mmin} + \frac{\tau_{mdiff}}{T_{adj} \left(1 + \exp\left(-\frac{V_{offmt} - V}{V_{slomt}}\right)\right)} \\ \tau_h &= \tau_{hmin} + \frac{\tau_{hdiff}}{T_{adj} \left(1 + \exp\left(-\frac{V_{offht} - V}{V_{sloht}}\right)\right)}\end{aligned}$$

$V_{offma} = -40.0$ mV, $V_{offmt} = -35.0$ mV, $V_{offha} = -90.0$ mV, $V_{offht} = -50.0$ mV, $V_{sloa} = 6.0$ mV, $V_{slomt} = 5.0$ mV, $V_{sloha} = 6.4$ mV, $V_{sloht} = 7.0$ mV, $\tau_{mmin} = 5.0$ ms, $\tau_{mdiff} = 20.0$ ms, $\tau_{hmin} = 20.0$ ms, $\tau_{hdiff} = 50.0$ ms, $\bar{g}_s = 0.00311$ S/cm², $\bar{g}_{a_1} = 0.000487$ S/cm², $N_m = 2$, $N_h = 1$

Small-conductance Ca²⁺-activated K⁺ current, I_{SK}

$$m_\infty = \frac{1}{1 + \left(\frac{[Ca^{2+}]_i}{c_{off}}\right)^{-c_{slo}}}$$

$c_{off} = 0.00043$ mM, $c_{slo} = 4.8$, $\bar{g}_s = 0.0479$ S/cm², $\bar{g}_{a_0} = 0.000231$ S/cm², $\bar{g}_{a_1} = 0.00365$ S/cm², $N_m = 1$, $N_h = 1$

Leak current, I_{leak}

$E = -90$ mV, $\bar{g}_s = 0.000078$ S/cm², $\bar{g}_{a_0} = 0.0000592$ S/cm², $\bar{g}_{a_1} = 0.0000675$ S/cm²,

$$\bar{g}_b = 0.0000256 \text{ S/cm}^2, N_m = 0, N_h = 0$$

Intracellular $[\text{Ca}^{2+}]$ dynamics

The intracellular Ca^{2+} concentration follows Equation 1 with the following model parameters: $\gamma_s = 0.0005$, $\gamma_{a_0} = 0.0347$, $\gamma_{a_1} = 0.0005$, $\tau_{\text{decay},s} = 488 \text{ ms}$, $\tau_{\text{decay},a_0} = 142 \text{ ms}$, $\tau_{\text{decay},a_1} = 95.4 \text{ ms}$, $d = 0.1 \text{ } \mu\text{m}$, $c_{\text{min}} = 10^{-4} \text{ mM}$

Temperature adjustment factor: $T_{\text{adj}} = 2.3^{\frac{34-21}{10}}$

Inhibitory interneurons Fast-spiking interneurons were also modelled as multi-compartment Hodgkin-Huxley type neurons. The model was taken from Vierling-Claassen et al. [103] which can be found in ModelDB (<https://senselab.med.yale.edu/modeldb/ShowModel?model=141273>). The evolution of its membrane potential over time is governed by the following differential equation

$$C_m \frac{\partial V}{\partial t} = I_{\text{Nat}} + I_K + I_l + I_{\text{axial}},$$

where the currents are modelled with the same formalisms as for the pyramidal cells. The two ionic currents I_{Nat} and I_K were modelled as in [57]. The basket cell model consisted of 16 compartments: the soma, and 15 dendritic compartments. The maximal conductance values \bar{g} were different for different compartments and are subindexed as follows: s refers to the soma, d to the dendrite (for details see [103]). All values not shown are set to 0.

Fast inactivating Na^+ current, I_{Nat}

$$\alpha_m = -\frac{1}{\tau_{\text{ma}}} \cdot \frac{V_{\text{offm}} - V}{1 - \exp\left(\frac{(V_{\text{offm}} - V)}{V_{\text{slo m}}}\right)}$$

$$\beta_m = \frac{1}{\tau_{\text{mb}}} \cdot \frac{V_{\text{offm}} - V}{1 - \exp\left(\frac{-(V_{\text{offm}} - V)}{V_{\text{slo m}}}\right)}$$

$$\alpha_h = \frac{1}{\tau_{\text{ha}}} \cdot \frac{V_{\text{offh}} - V}{1 - \exp\left(\frac{(V_{\text{offh}} - V)}{V_{\text{slo h}}}\right)}$$

$$\beta_h = -\frac{1}{\tau_{\text{hb}}} \cdot \frac{V_{\text{offh}} - V}{1 - \exp\left(\frac{-(V_{\text{offh}} - V)}{V_{\text{slo h}}}\right)}$$

$$m_{\infty} = \frac{\alpha_m}{\alpha_m + \beta_m}$$

$$h_{\infty} = \frac{\alpha_h}{\alpha_h + \beta_h}$$

$$\tau_m = \frac{1}{T_{\text{adj}}(\alpha_m + \beta_m)}$$

$$\tau_h = \frac{1}{T_{\text{adj}}(\alpha_h + \beta_h)}$$

$V_{\text{offm}} = -25$ mV, $V_{\text{offh}} = -25$ mV, $V_{\text{slohm}} = 9.0$ mV, $V_{\text{sloh}} = 9.0$ mV, $\tau_{\text{ma}} = 5.49$ ms, $\tau_{\text{mb}} = 8.06$ ms, $\tau_{\text{ha}} = 41.67$ ms, $\tau_{\text{hb}} = 109.89$ ms, $\bar{g}_s = 0.06$ S/cm², $\bar{g}_d = 0.035$ S/cm², $N_m = 3$, $N_h = 1$

K⁺ current, I_K

$$\alpha_m = -\frac{1}{\tau_{\text{ma}}} \cdot \frac{V_{\text{offm}} - V}{1 - \exp\left(\frac{V_{\text{offm}} - V}{V_{\text{slohm}}}\right)}$$

$$\beta_m = \frac{1}{\tau_{\text{mb}}} \cdot \frac{V_{\text{offm}} - V}{1 - \exp\left(\frac{-(V_{\text{offm}} - V)}{V_{\text{slohm}}}\right)}$$

$$m_\infty = \frac{\alpha_m}{\alpha_m + \beta_m}$$

$$\tau_m = \frac{1}{T_{\text{adj}}(\alpha_m + \beta_m)}$$

$V_{\text{offm}} = -25$ mV, $V_{\text{offh}} = -25$ mV, $\tau_{\text{ma}} = 50.0$ ms, $\tau_{\text{mb}} = 500.0$ ms, $\bar{g}_s = 0.05$ S/cm², $\bar{g}_d = 0.035$ S/cm², $N_m = 1$.

Leak current, I_{leak}

$E = -73$ mV, $\bar{g}_s = 0.000012$ S/cm², $N_m = 0$, $N_h = 0$

For the interneurons, Na⁺ and K⁺ have a fixed reversal potential of $E_{\text{Na}} = 50$ mV and $E_{\text{K}} = -85$ mV.

Description of the Network Model

The network model consisted of 256 layer 5 pyramidal cells and 64 fast-spiking inhibitory interneurons and was implemented in NEURON [38] using the NetPyNE interface [25]. Pyramidal cells were connected to each other randomly with a probability of 0.06 (using AMPA and NMDA synaptic receptors). AMPA synaptic receptors were modelled as a double exponential function with a rise time of 0.1 ms and decay time of 3.0 ms. NMDA receptors were modelled as a first-order model including different binding and unbinding rates together with a magnesium block [20, 21]. AMPA receptors had a maximal conductance of 0.0012 nS and the NMDA receptors of 0.0006 nS, reflecting the higher contribution of AMPA receptors (see [75, 89]). Pyramidal cells were connected to inhibitory interneurons with a probability of 0.43, again using AMPA and NMDA receptors. However, for this connection type the maximal conductances for AMPA receptors, 0.0012 nS, was substantially higher than for NMDA, 0.00013, reflecting the minor role NMDAergic activation plays in the recruitment of PV⁺ interneurons [73, 89, 31]. Inhibitory interneurons formed connections with pyramidal cells and with themselves with a probability of 0.44 and 0.51, respectively. Inhibitory connections were realised using GABA_A receptors, which were also modelled as double exponential functions with a rise time of 0.5 ms and a decay time of 8.0 ms (except for the

condition of prolonged inhibitory decay times, where the decay time is increased to 25.0 ms). Maximal conductance values for inhibitory connections to pyramidal cells was 0.035 nS and to themselves 0.023 nS.

The network received two types of input: 1) Poissonian noise reflecting random cortical and subcortical background activity and 2) periodic input representing auditory steady-state stimuli at gamma frequency (operationalized at 40 Hz in this study). For the random background noise we utilized NEURON's built-in *NetStim* function with $rate = 200$, $noise = 1.0$ and $start = 500$. The parameters result in a noise stimulus reflecting a background rate of 200 Hz, that is completely random (i.e. Poissonian) which starts after 500 ms and lasts until the end of the simulation. The background *NetStim* element was connected to the pyramidal cells with a weight of 0.0325 and to the interneurons with a weight of 0.002. This setting resulted in asynchronous irregular firing of both cell populations at rates of 7.81 ± 0.07 Hz for the pyramidal cells and 6.18 ± 0.26 Hz for the inhibitory neurons, which is close to the reported spontaneous activity of ~ 4.9 Hz reported for auditory cortex [42]. The periodic input resembling auditory click-train stimuli was also realised with NEURON's built-in *NetStim* function with $rate = 40.0$, $noise = 0.0$ and $start = 1000$. These parameters result in a fully periodic stimulus with an inter-event interval of 25 ms, starting after 1000 ms and ending with the end of the simulation. In summary, the network was allowed to settle for the first 500 ms, then background input was switched on and after another 500 ms stimulus input was switched on. We simulated a total of 2000 ms and used only the last 1000 ms for the analysis of steady-state entrainment.

In our network model both, the background noise and the specific connections between neurons, are based on stochastic processes. To ensure that our findings do not depend on the specific instantiations of these stochastic processes, we always performed multiple simulations with different seeds for the random number generators underlying the processes. Specifically, we simulated 20 *subjects* by choosing 20 different seeds for the random number generator underlying the formation of synaptic connections, i.e. each *subject* had the same connectivity throughout all simulations. For each *subject* we then performed 10 *trials* by choosing 10 different random seeds for the random number generator underlying the background noise. In total, we performed 200 simulations for each network configuration, with the different network configurations being 1) the control network, 2) a single model variant, 3) a combination of different model variants, and 4) a synaptic alteration (either a reduced g_{max} or a increased τ_{decay} at GABAergic synapses).

Modelling SNP-like Genetic Variants

The single cell models include Hodgkin-Huxley type description for channel activation and inactivation, and hence, changes related to certain ion-channel-encoding gene variants that have been observed in experiments can be directly attributed to a change of one or more parameters of these models. However, we only modelled changes to the pyramidal cells, which have been found to be strongly implicated by many schizophrenia-associated gene sets [88].

Here, we restricted ourselves to the following set of ion channel-encoding genes:

CACNA1C, CACNA1D, CACNB2, CACNA1I, SCN1A and, HCN1. For details on the selected genes and variants see also [60, 63]. Supplementary table S1 gives all details of the different variants of these genes and in Supplementary tables S2 and S3 all details on how they were integrated into the pyramidal cell model are shown.

Extraction of Functional Genomics Studies The modelling of SNP-like genetic variants is based on an extensive search of the literature on how the genes CACNA1C, CACNA1D, CACNB2, CACNA1I, SCN1A, and HCN1 change the dynamics of the underlying ion channel. We included studies performed in different animal species and different cell types because of a current lack of data for a single animal and tissue type. Our inclusion criteria were as follows:

- The study applied a genetic variant of one of the genes of interest.
- The features of the variant-expressing cell were investigated using electrophysiology or Ca^{2+} imaging.
- The change from the control cell behaviour to the variant cell behaviour could be meaningfully translated to the layer 5 pyramidal cell model.
- The observed effect of the gene variant was not purely an effect of ion channel density or expression level.

We applied the last criterion mainly because there are numerous ways that might influence such an effect [84], while the changes of ion channel dynamics are supposedly more directly dependent on the genetic encoding. Supplementary table S1 lists all the pertinent data from the included studies [50, 19, 40, 91, 94, 93, 5, 112, 79, 80, 3, 54, 17, 66, 55, 43, 14, 102, 105, 13, 65, 44, 53, 108].

In studies that reported the effects of several variant types, the ranges of possible effects are considered. In case variants considered in such studies yielded positive and negative effects, both were included, however, if the reported endpoints of the ranges were too close to the control value (i.e. less than 1 mV or less than 10% of the distance between control value and the other endpoint), only the larger deviation was considered.

Scaling the Gene Variants Due to the polygenic nature of psychiatric disorders such as SCZ, it seems reasonable to assume that the disorder is not caused by a single SCZ-associated SNP but rather by a combination of sufficiently many (cf. [52]). Therefore, we followed earlier approaches [60, 63] to scale down the effect of single variants. We proceeded as follows. If a variant of the model (as shown in Supplementary Table 1) changed the response of the neuron too strongly, we scaled down the effect of the genetic variant until the differences between control and variant model neuron were within a given bound. Specifically, down-scaling was performed until the following five criteria were fulfilled ([60, 63]):

1. Exactly 4 spikes were induced as a response to current injection of 0.696 nA for the duration of 150 ms (square pulse),
2. exactly 1 spike was induced as a response to a distal (distance from soma = 620 μm) synaptic conductance (alpha-shaped with a time constant of 5 ms and a maximum amplitude of 0.0612 μS),
3. the neuron responded with exactly 2 spikes to a combined stimulus of a somatic current injection (1.137 nA, 10 ms, square pulse) and a distal synaptic conductance (alpha-shaped with a time constant of 5 ms and a maximum amplitude of 0.100 μS , applied 5 ms after the somatic pulse),
4. the integrated difference between the f-I curves of the variant model neuron and the control model neuron did not exceed 10% of the integral of the control neuron f-I curve, and
5. the membrane-potential limit cycle should not be too different from the control neuron limit cycle ($d_{cc}(lc1, lc2) \leq 600$, see [60] for the definition of the metric d_{cc}).

Here, the first three conditions constrain the magnitudes of transient responses of the neuron model and the amplitudes were chosen to guarantee the largest stability with respect to the response of the model using default parameters (stable here, means that an equal change in current amplitude on a logarithmic scale is needed to produce one action potential more or one action potential less).

In the case of a violation of one or more conditions, the effect of the variant on the parameters of the model was down-scaled to a fraction $c < 1$, where the violation is observed for the first time. The fraction for each considered variant can be found in Supplementary table S2. If however an unscaled variant did not violate the conditions 1–5, we explored threshold effects up to twice the original effect, i.e., $c \leq 2$.

Since the parameters of the neuron model subject to change by the variants were very diverse, i.e. had various different roles and dimensions (mV, mM, ms, etc.), we adopted a careful scaling strategy. Parameters that could take positive and negative values were scaled linearly with the scaling factor c . Parameters that were exclusively non-negative, on the other hand, were scaled logarithmically with c . In detail, this meant that the difference in offset potentials (V_{offm} , V_{offh}) between the control model neuron and the variant model neuron, which are described by an additive term (i.e.

$\pm x m V$), was multiplied by the down-scaling parameter c . However, the change in all other parameters, which were described by multiplicative terms, was exponentiated by the down-scaling factor c . This resulted in a continuous change of parameters for $c \in [0, 1]$, which is also directly extendable to down-scaling factors > 1 .

Table 1: Overview over the genetic variants. The first column of the table shows the gene whose variant was studied in the named reference. Columns two and three show the current species and the affected model parameters, *offm* and *offh* representing the mid-points of activation and inactivation, respectively, *slo*m and *slo*h their respective slopes, and *taum* and *tauh* their respective time constants. Multiple parameter changes in a single row are separated by a semicolon . The parameter names may refer to multiple model parameters: For example, for HVA Ca^{2+} currents, a change in *offm* means a concurrent change in parameters V_{offma} and V_{offmb} , while *offh* refers to parameters V_{offha} and V_{offhb} , *slo*m to V_{sloma} and $V_{\text{slo mb}}$, *slo*h to V_{sloha} and $V_{\text{slo hb}}$, *taum* to τ_{ma} and τ_{mb} , and *tauh* to τ_{ha} and τ_{hb} . See model equations in Supplementary material for further details. The fourth column shows both direction and magnitude of the effect, where $\pm x$ mV refers to a change of the mid-point of the (in)activation curve by an absolute number, and $\pm x\%$ refers to a percentage change in the underlying quantity. In many cases, studies considered more than one variant and here, they are categorized by the type of the variant where necessary (e.g., in [50] several variants of four loci, of which three were in pore-lining IS6 segment and two in bundle-crossing region of segment IIS6, were considered, and the variants are here categorized according to the segment they acted on). Column five gives the type of variant used, while the last column names the cell type under consideration.

Gene	Current	Parameter	Effect	Type of variant	Cell type
CACNA1C [50]	I_{CaHVA}	offm; offh	-25.9..-1.4mV; -27.0..-3.8mV	L429T, L434T, S435T, S435A, S435P	TSA201
CACNA1C [50]	I_{CaHVA}	offm; offh	-37.3..-9.7mV; -30.0..-11.8mV	L779T, I781T, I781P	TSA201
CACNA1C [19]	I_{CaHVA}	offm; offh; slo; sloh	-31.4..+7.0mV; -28.5..+16.3mV -15..+45%; -28%..+38%	G432X, A780X, G1193X, A1503X	TSA201
CACNA1C [40]	I_{CaHVA}	offm; slo	-38.5..+12.9mV -54..+56%	I781X, C769P, G770P, N771P, I773P F778P, L779P, A780P, A782P, V783P	TSA201
CACNA1C [91]	I_{CaHVA}	offm; offh; slo	-27.8..+8.7mV; -19.1..+4.7mV -11%..+14%	I781T, N785A, N785G, N785L	TSA201
CACNA1C [94]	I_{CaHVA}	offm; offh; sloh	-11.2..+1.0mV; -3.1..-0.3mV +3%..+24%	Splice variants a1C77-A, -B, -C, and -D	TSA201
CACNA1D [93] [5]	I_{CaHVA}	offm; offh; slo; sloh; tauh	-10.9..-8.5mV; -3.0..+3.5mV -27..-13%; -12..-19% +25%	Splice variant 42A	TSA201 / HEK293
CACNA1D [93] [5]	I_{CaHVA}	offm; offh; slo; sloh; tauh	-10.6..+3.4mV; -5.3..+1.2mV -20..+12%; -34..-8% -28%	Splice variant 43S	TSA201 / HEK293
CACNA1D [112] [79]	I_{CaHVA}	offm; slo; tauh	+3.5..+6.6mV -25..+19%; -50..+12%	Homozygous knockout	AV-node / chromaffin cells
CACNA1D [80]	I_{CaHVA}	offm; offh; slo; sloh	-9.8mV; -15.4mV -20%; sloh	A749G	TSA201
CACNA1D [3]	I_{CaHVA}	offm; offh; slo; sloh; tauh	-24.2..+6.1mV; -14.5..-3.6mV -30..+24%; -28%..+28% +43%..+252%	V259D, I750M, P1336R	TSA201
CACNA1D [54]	I_{CaHVA}	offm; slo; tauh	-17.8..-13.1mV -19..-0%; -23%..+31%	rCav1.3scg variant and mutants 7M2K, S244G, V1104A, and A2075V	TSA201
CACNB2 [17]	I_{CaHVA}	offh; sloh	-5.2mV; -31%	T111	TSA201
CACNB2 [66]	I_{CaHVA}	taum	+70%	A1B2 vs A1 alone	HEK293
CACNB2 [55]	I_{CaHVA}	offm; offh; tauh; tauh	-4.9..+4.9mV; -5.1..+5.1mV -40%..+68%; -40%..+66% +26%	Splice variants N1, N3, N4, and N5	HEK293
CACNB2 [43]	I_{CaHVA}	tauh	+26%	D601E	TSA201
CACNA1I [74]	I_{CaLVA}	offm; offh; tauh; tauh	-0.2..+1.3mV; -0.5..+1.6mV -13..+45%; -20..+8%	Alternative splicing of exons 9 and 33	HEK293
CACNA1I [30]	I_{CaLVA}	offm; offh; slo; sloh; tauh; tauh	-4.3..-1.2mV; -4.4..-1.9mV +5..+14%; -11%..+4% -47%..-15%; -54%..+1%	Truncated cDNAs L4, L6 and L9	HEK293
CACNA1I [30]	I_{CaLVA}	offm; offh; slo; sloh; tauh; tauh	-4.3..-1.2mV; -4.4..-1.9mV +5..+14%; -11%..+4% -47%..-15%; -54%..+1%	Truncated cDNAs L4, L6 and L9	HEK293
SCN1A [14]	I_{Nat}	offm; offh; slo; sloh	-0.3mV; +5.0mV +15%; +23%	Q1489K	Cultured neocortical cells
SCN1A [102]	I_{Nat}	offm; offh; slo; sloh	+2.8mV; +6.3..+9.6mV +1.6%; +4.2%	L1649Q	TSA201
SCN1A [105]	I_{Nat}	offm; offh; slo; sloh; tauh	-4.0mV; -5.8mV -8%; +13%; +43..+47%	R859H	TSA201
SCN1A [105]	I_{Nat}	offm; offh; slo; sloh; tauh	-8.1mV; +2.2mV -3%; -3%; +26..+59%	R865G	TSA201
SCN1A [13]	I_{Nat}	offm; slo; tauh	+6.0mV; +16%; +29%	T1174S	TSA201
SCN1A [65]	I_{Nat}	offm; offh; slo; sloh	+10.0mV; -0.6mV +15%; +14%	M145T	TSA201
HCN1 [44]	I_h	offh; sloh	-2.1..-26.5mV; -12..-36%	D135W, D135H, D135N	HEK293
HCN1 [53]	I_h	offh; sloh	-25.9..+17.7mV -40..+3%	E229A, K230A, G231A, M232A, D233A S234A, E235G, V236A, Y237A, EVY235-237DDD	Oocytes
HCN1 [108]	I_h	offh; tauh	+2.4..+3.9mV; -12..-0%	WAG-HCN1, WAG-HCN1 + HCN1 co-expression	Oocytes

Table 2: The effects of the genetic variants on model parameters. The first column names the gene in which the gene variant was analyzed and the study where the effects were reported. The second column shows the effect of the variant on the model parameters, here *offm* and *offh* represent the mid-points of activation and inactivation, respectively, *sлом* and *sloh* their respective slopes, and *taum* and *tauh* their respective time constants. The third column shows the down-scaling parameter resulting from the scaling procedure outlined above. The horizontal separation of rows refers to the different entries of Supplementary table 1: If the corresponding study showed a large range of effects on single model parameters, the endpoints of such ranges are here treated as separate variants and are therefore downscaled independently of each other. The variants shown in Figures 3, 4 and 5 are marked with an asterisk.

Gene	Parameter changes	Scaling parameter
<i>CACNA1C</i> [50]	$V_{offm^*,CaHVA} : -25.9 \text{ mV}; V_{offh^*,CaHVA} : -27.0 \text{ mV}$	$c_{thresh} = 0.066$
<i>CACNA1C</i> [50]	$V_{offm^*,CaHVA} : -37.3 \text{ mV}; V_{offh^*,CaHVA} : -30.0 \text{ mV}$	$c_{thresh} = 0.042$
<i>CACNA1C</i> [19]	$V_{offm^*,CaHVA} : -31.4 \text{ mV}; V_{sлом^*,CaHVA} : *0.85; V_{offh^*,CaHVA} : -28.5 \text{ mV}; V_{sloh^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : +7.0 \text{ mV}; V_{sлом^*,CaHVA} : *0.85; V_{offh^*,CaHVA} : -28.5 \text{ mV}; V_{sloh^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : -31.4 \text{ mV}; V_{sлом^*,CaHVA} : *1.45; V_{offh^*,CaHVA} : -28.5 \text{ mV}; V_{sloh^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : +7.0 \text{ mV}; V_{sлом^*,CaHVA} : *1.45; V_{offh^*,CaHVA} : -28.5 \text{ mV}; V_{sloh^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : -31.4 \text{ mV}; V_{sлом^*,CaHVA} : *0.85; V_{offh^*,CaHVA} : +16.3 \text{ mV}; V_{sloh^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : +7.0 \text{ mV}; V_{sлом^*,CaHVA} : *0.85; V_{offh^*,CaHVA} : +16.3 \text{ mV}; V_{sloh^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : -31.4 \text{ mV}; V_{sлом^*,CaHVA} : *1.45; V_{offh^*,CaHVA} : +16.3 \text{ mV}; V_{sloh^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : +7.0 \text{ mV}; V_{sлом^*,CaHVA} : *1.45; V_{offh^*,CaHVA} : +16.3 \text{ mV}; V_{sloh^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : -31.4 \text{ mV}; V_{sлом^*,CaHVA} : *0.85; V_{offh^*,CaHVA} : -28.5 \text{ mV}; V_{sloh^*,CaHVA} : *1.38$ $V_{offm^*,CaHVA} : +7.0 \text{ mV}; V_{sлом^*,CaHVA} : *0.85; V_{offh^*,CaHVA} : -28.5 \text{ mV}; V_{sloh^*,CaHVA} : *1.38$ $V_{offm^*,CaHVA} : -31.4 \text{ mV}; V_{sлом^*,CaHVA} : *1.45; V_{offh^*,CaHVA} : -28.5 \text{ mV}; V_{sloh^*,CaHVA} : *1.38$ $V_{offm^*,CaHVA} : +7.0 \text{ mV}; V_{sлом^*,CaHVA} : *1.45; V_{offh^*,CaHVA} : -28.5 \text{ mV}; V_{sloh^*,CaHVA} : *1.38$ $V_{offm^*,CaHVA} : -31.4 \text{ mV}; V_{sлом^*,CaHVA} : *0.85; V_{offh^*,CaHVA} : +16.3 \text{ mV}; V_{sloh^*,CaHVA} : *1.38$ $V_{offm^*,CaHVA} : +7.0 \text{ mV}; V_{sлом^*,CaHVA} : *0.85; V_{offh^*,CaHVA} : +16.3 \text{ mV}; V_{sloh^*,CaHVA} : *1.38$ $V_{offm^*,CaHVA} : -31.4 \text{ mV}; V_{sлом^*,CaHVA} : *1.45; V_{offh^*,CaHVA} : +16.3 \text{ mV}; V_{sloh^*,CaHVA} : *1.38$ $V_{offm^*,CaHVA} : +7.0 \text{ mV}; V_{sлом^*,CaHVA} : *1.45; V_{offh^*,CaHVA} : +16.3 \text{ mV}; V_{sloh^*,CaHVA} : *1.38$	$c_{thresh} = 0.043$ $c_{thresh} = 0.101$ $c_{thresh} = 0.049$ $c_{thresh} = 0.76$ $c_{thresh} = 0.076$ $c_{thresh} = 0.693$ $c_{thresh} = 0.034$ $c_{thresh} = 0.359$ $c_{thresh} = 0.059$ $c_{thresh} = 0.103$ $c_{thresh} = 0.049$ $c_{thresh} = 0.176$ $c_{thresh} = 0.038$ $c_{thresh} = 0.113$
<i>CACNA1C</i> [40]	$V_{offm^*,CaHVA} : -38.5 \text{ mV}; V_{sлом^*,CaHVA} : *0.46$ $V_{offm^*,CaHVA} : +12.9 \text{ mV}; V_{sлом^*,CaHVA} : *0.46$	$c_{thresh} = 0.028$ $c_{thresh} = 0.123$
<i>CACNA1C</i> [91]	$V_{offm^*,CaHVA} : -27.8 \text{ mV}; V_{sлом^*,CaHVA} : *0.89; V_{offh^*,CaHVA} : -19.1 \text{ mV}$ $V_{offm^*,CaHVA} : +8.7 \text{ mV}; V_{sлом^*,CaHVA} : *0.89; V_{offh^*,CaHVA} : -19.1 \text{ mV}$ $V_{offm^*,CaHVA} : -27.8 \text{ mV}; V_{sлом^*,CaHVA} : *1.14; V_{offh^*,CaHVA} : -19.1 \text{ mV}$ $V_{offm^*,CaHVA} : +8.7 \text{ mV}; V_{sлом^*,CaHVA} : *1.14; V_{offh^*,CaHVA} : -19.1 \text{ mV}$ $V_{offm^*,CaHVA} : -27.8 \text{ mV}; V_{sлом^*,CaHVA} : *0.89; V_{offh^*,CaHVA} : +4.7 \text{ mV}$ $V_{offm^*,CaHVA} : +8.7 \text{ mV}; V_{sлом^*,CaHVA} : *0.89; V_{offh^*,CaHVA} : +4.7 \text{ mV}$ $V_{offm^*,CaHVA} : -27.8 \text{ mV}; V_{sлом^*,CaHVA} : *1.14; V_{offh^*,CaHVA} : +4.7 \text{ mV}$ $V_{offm^*,CaHVA} : +8.7 \text{ mV}; V_{sлом^*,CaHVA} : *1.14; V_{offh^*,CaHVA} : +4.7 \text{ mV}$	$c_{thresh} = 0.052$ $c_{thresh} = 0.077$ $c_{thresh} = 0.057$ $c_{thresh} = 0.069$ $c_{thresh} = 0.042$ $c_{thresh} = 0.145$ $c_{thresh} = 0.044$ $c_{thresh} = 0.119$
<i>CACNA1C</i> [94]	$V_{offm^*,CaHVA} : -11.2 \text{ mV}; V_{offh^*,CaHVA} : -3.1 \text{ mV}; V_{sloh^*,CaHVA} : *1.24$ $V_{offm^*,CaHVA} : +1.0 \text{ mV}; V_{offh^*,CaHVA} : -3.1 \text{ mV}; V_{sloh^*,CaHVA} : *1.24$	$c_{thresh} = 0.157$ $c_{thresh} = 0.236$
<i>CACNA1D</i> [93], [5]	$V_{offm^*,CaHVA} : -10.9 \text{ mV}; V_{sлом^*,CaHVA} : *0.73; V_{offh^*,CaHVA} : -3.0 \text{ mV}; V_{sloh^*,CaHVA} : *0.81;$ $\tau_{h^*,CaHVA} : *1.25$ $V_{offm^*,CaHVA} : -10.9 \text{ mV}; V_{sлом^*,CaHVA} : *0.73; V_{offh^*,CaHVA} : +3.5 \text{ mV}; V_{sloh^*,CaHVA} : *0.81;$ $\tau_{h^*,CaHVA} : *1.25$	$c_{thresh} = 0.083$ $c_{thresh} = 0.075$
<i>CACNA1D</i> [93], [5]	$V_{offm^*,CaHVA} : -10.6 \text{ mV}; V_{sлом^*,CaHVA} : *0.8; V_{offh^*,CaHVA} : -5.3 \text{ mV}; V_{sloh^*,CaHVA} : *0.66;$ $\tau_{h^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : +3.4 \text{ mV}; V_{sлом^*,CaHVA} : *0.8; V_{offh^*,CaHVA} : -5.3 \text{ mV}; V_{sloh^*,CaHVA} : *0.66;$ $\tau_{h^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : -10.6 \text{ mV}; V_{sлом^*,CaHVA} : *1.12; V_{offh^*,CaHVA} : -5.3 \text{ mV}; V_{sloh^*,CaHVA} : *0.66;$ $\tau_{h^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : +3.4 \text{ mV}; V_{sлом^*,CaHVA} : *1.12; V_{offh^*,CaHVA} : -5.3 \text{ mV}; V_{sloh^*,CaHVA} : *0.66;$ $\tau_{h^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : -10.6 \text{ mV}; V_{sлом^*,CaHVA} : *0.8; V_{offh^*,CaHVA} : +1.2 \text{ mV}; V_{sloh^*,CaHVA} : *0.66;$ $\tau_{h^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : +3.4 \text{ mV}; V_{sлом^*,CaHVA} : *0.8; V_{offh^*,CaHVA} : +1.2 \text{ mV}; V_{sloh^*,CaHVA} : *0.66;$ $\tau_{h^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : -10.6 \text{ mV}; V_{sлом^*,CaHVA} : *1.12; V_{offh^*,CaHVA} : +1.2 \text{ mV}; V_{sloh^*,CaHVA} : *0.66;$ $\tau_{h^*,CaHVA} : *0.72$ $V_{offm^*,CaHVA} : +3.4 \text{ mV}; V_{sлом^*,CaHVA} : *1.12; V_{offh^*,CaHVA} : +1.2 \text{ mV}; V_{sloh^*,CaHVA} : *0.66;$ $\tau_{h^*,CaHVA} : *0.72$	$c_{thresh} = 0.080$ $c_{thresh} = 1.962$ $c_{thresh} = 0.094$ $c_{thresh} = 0.905$ $c_{thresh} = 0.072$ $c_{thresh} = 0.386$ $c_{thresh} = 0.083$ $c_{thresh} = 1.117$
<i>CACNA1D</i> [112], [79]	$V_{offm^*,CaHVA} : +6.6 \text{ mV}; V_{sлом^*,CaHVA} : *0.75; \tau_{h^*,CaHVA} : *0.5$ $V_{offm^*,CaHVA} : +6.6 \text{ mV}; V_{sлом^*,CaHVA} : *1.19; \tau_{h^*,CaHVA} : *0.5$ $V_{offm^*,CaHVA} : +6.6 \text{ mV}; V_{sлом^*,CaHVA} : *0.75; \tau_{h^*,CaHVA} : *1.12$ $V_{offm^*,CaHVA} : +6.6 \text{ mV}; V_{sлом^*,CaHVA} : *1.19; \tau_{h^*,CaHVA} : *1.12$	$c_{thresh} = 0.190$ $c_{thresh} = 0.123$ $c_{thresh} = 0.209$ $c_{thresh} = 0.130$
<i>CACNA1D</i> [80]	$V_{offm^*,CaHVA} : -9.8 \text{ mV}; V_{sлом^*,CaHVA} : *0.8; V_{offh^*,CaHVA} : -15.4 \text{ mV}; V_{sloh^*,CaHVA} : *1.05$	$c_{thresh} = 0.181$
<i>CACNA1D</i> [3]	$V_{offm^*,CaHVA} : -24.2 \text{ mV}; V_{sлом^*,CaHVA} : *0.7; V_{offh^*,CaHVA} : -14.5 \text{ mV}; V_{sloh^*,CaHVA} : *0.72;$ $\tau_{h^*,CaHVA} : *3.52$ $V_{offm^*,CaHVA} : +6.1 \text{ mV}; V_{sлом^*,CaHVA} : *0.7; V_{offh^*,CaHVA} : -14.5 \text{ mV}; V_{sloh^*,CaHVA} : *0.72;$ $\tau_{h^*,CaHVA} : *3.52$ $V_{offm^*,CaHVA} : -24.2 \text{ mV}; V_{sлом^*,CaHVA} : *1.24; V_{offh^*,CaHVA} : -14.5 \text{ mV}; V_{sloh^*,CaHVA} : *0.72;$ $\tau_{h^*,CaHVA} : *3.52$ $V_{offm^*,CaHVA} : +6.1 \text{ mV}; V_{sлом^*,CaHVA} : *1.24; V_{offh^*,CaHVA} : -14.5 \text{ mV}; V_{sloh^*,CaHVA} : *0.72;$ $\tau_{h^*,CaHVA} : *3.52$ $V_{offm^*,CaHVA} : -24.2 \text{ mV}; V_{sлом^*,CaHVA} : *0.7; V_{offh^*,CaHVA} : -14.5 \text{ mV}; V_{sloh^*,CaHVA} : *1.28;$ $\tau_{h^*,CaHVA} : *3.52$ $V_{offm^*,CaHVA} : +6.1 \text{ mV}; V_{sлом^*,CaHVA} : *0.7; V_{offh^*,CaHVA} : -14.5 \text{ mV}; V_{sloh^*,CaHVA} : *1.28;$ $\tau_{h^*,CaHVA} : *3.52$ $V_{offm^*,CaHVA} : -24.2 \text{ mV}; V_{sлом^*,CaHVA} : *1.24; V_{offh^*,CaHVA} : -14.5 \text{ mV}; V_{sloh^*,CaHVA} : *1.28;$ $\tau_{h^*,CaHVA} : *3.52$ $V_{offm^*,CaHVA} : +6.1 \text{ mV}; V_{sлом^*,CaHVA} : *1.24; V_{offh^*,CaHVA} : -14.5 \text{ mV}; V_{sloh^*,CaHVA} : *1.28;$ $\tau_{h^*,CaHVA} : *3.52$	$c_{thresh} = 0.045$ $c_{thresh} = 0.318$ $c_{thresh} = 0.053$ $c_{thresh} = 0.152$ $c_{thresh} = 0.059$ $c_{thresh} = 0.105$ $c_{thresh} = 0.074$ $c_{thresh} = 0.076$

Table 3: Table 2 continued.

Gene	Parameter changes	Scaling parameter
<i>CACNA1D</i> [54]	$V_{\text{offm}^*,\text{CaHVA}}: -17.8 \text{ mV}; V_{\text{slo}m^*,\text{CaHVA}}: *0.81; \tau_{\text{h}^*,\text{CaHVA}}: *0.77$	$c_{\text{thresh}} = 0.065$
	$V_{\text{offm}^*,\text{CaHVA}}: -17.8 \text{ mV}; V_{\text{slo}m^*,\text{CaHVA}}: *0.81; \tau_{\text{h}^*,\text{CaHVA}}: *1.31$	$c_{\text{thresh}} = 0.063$
<i>CACNB2</i> [17]	$V_{\text{offh}^*,\text{CaHVA}}: -5.2 \text{ mV}; V_{\text{slo}h^*,\text{CaHVA}}: *0.69$	$c_{\text{thresh}} = 0.381$
<i>CACNB2</i> [66]	$\tau_{\text{h}^*,\text{CaHVA}}: *1.7$	$c_{\text{thresh}} = 2.000$
<i>CACNB2</i> [55]	$V_{\text{offm}^*,\text{CaHVA}}: -4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: -5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *0.6; \tau_{\text{h}^*,\text{CaHVA}}: *0.6$	$c_{\text{thresh}} = 0.194$
	$V_{\text{offm}^*,\text{CaHVA}}: +4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: -5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *0.6; \tau_{\text{h}^*,\text{CaHVA}}: *0.6$	$c_{\text{thresh}} = 0.157$
	$V_{\text{offm}^*,\text{CaHVA}}: -4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: +5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *0.6; \tau_{\text{h}^*,\text{CaHVA}}: *0.6$	$c_{\text{thresh}} = 0.122$
	$V_{\text{offm}^*,\text{CaHVA}}: +4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: +5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *0.6; \tau_{\text{h}^*,\text{CaHVA}}: *0.6$	$c_{\text{thresh}} = 1.101$
	$V_{\text{offm}^*,\text{CaHVA}}: -4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: -5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *1.68; \tau_{\text{h}^*,\text{CaHVA}}: *0.6$	$c_{\text{thresh}} = 0.814$
	$V_{\text{offm}^*,\text{CaHVA}}: +4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: -5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *1.68; \tau_{\text{h}^*,\text{CaHVA}}: *0.6$	$c_{\text{thresh}} = 0.146$
	$V_{\text{offm}^*,\text{CaHVA}}: -4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: +5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *1.68; \tau_{\text{h}^*,\text{CaHVA}}: *0.6$	$c_{\text{thresh}} = 0.517$
	$V_{\text{offm}^*,\text{CaHVA}}: +4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: +5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *1.68; \tau_{\text{h}^*,\text{CaHVA}}: *0.6$	$c_{\text{thresh}} = 0.204$
	$V_{\text{offm}^*,\text{CaHVA}}: -4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: -5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *0.6; \tau_{\text{h}^*,\text{CaHVA}}: *1.66$	$c_{\text{thresh}} = 0.285$
	$V_{\text{offm}^*,\text{CaHVA}}: +4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: -5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *0.6; \tau_{\text{h}^*,\text{CaHVA}}: *1.66$	$c_{\text{thresh}} = 0.511$
	$V_{\text{offm}^*,\text{CaHVA}}: -4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: +5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *0.6; \tau_{\text{h}^*,\text{CaHVA}}: *1.66$	$c_{\text{thresh}} = 0.197$
	$V_{\text{offm}^*,\text{CaHVA}}: +4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: +5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *0.6; \tau_{\text{h}^*,\text{CaHVA}}: *1.66$	$c_{\text{thresh}} = 1.687$
	$V_{\text{offm}^*,\text{CaHVA}}: -4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: -5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *1.68; \tau_{\text{h}^*,\text{CaHVA}}: *1.66$	$c_{\text{thresh}} = 0.707$
	$V_{\text{offm}^*,\text{CaHVA}}: +4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: -5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *1.68; \tau_{\text{h}^*,\text{CaHVA}}: *1.66$	$c_{\text{thresh}} = 0.156$
	$V_{\text{offm}^*,\text{CaHVA}}: -4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: +5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *1.68; \tau_{\text{h}^*,\text{CaHVA}}: *1.66$	$c_{\text{thresh}} = 0.460$
	$V_{\text{offm}^*,\text{CaHVA}}: +4.9 \text{ mV}; V_{\text{offh}^*,\text{CaHVA}}: +5.1 \text{ mV}; \tau_{\text{m}^*,\text{CaHVA}}: *1.68; \tau_{\text{h}^*,\text{CaHVA}}: *1.66$	$c_{\text{thresh}} = 0.218$
<i>CACNB2</i> [43]	$\tau_{\text{h}^*,\text{CaHVA}}: *1.26$	$c_{\text{thresh}} = 2.000$
<i>CACNA1I</i> [74]	$V_{\text{offma},\text{CaLVA}}: +1.3 \text{ mV}; V_{\text{offha},\text{CaLVA}}: +1.6 \text{ mV}; \tau_{\text{m}^*,\text{CaLVA}}: *0.87; \tau_{\text{h}^*,\text{CaLVA}}: *0.8$	$c_{\text{thresh}} = 2.000$
	$V_{\text{offma},\text{CaLVA}}: +1.3 \text{ mV}; V_{\text{offha},\text{CaLVA}}: +1.6 \text{ mV}; \tau_{\text{m}^*,\text{CaLVA}}: *1.45; \tau_{\text{h}^*,\text{CaLVA}}: *0.8$	$c_{\text{thresh}} = 2.000$ *
<i>CACNA1I</i> [30]	$V_{\text{offma},\text{CaLVA}}: -4.3 \text{ mV}; V_{\text{sloma},\text{CaLVA}}: *1.14; V_{\text{offha},\text{CaLVA}}: -4.4 \text{ mV}; V_{\text{sloha},\text{CaLVA}}: *0.89;$ $\tau_{\text{m}^*,\text{CaLVA}}: *0.53; \tau_{\text{h}^*,\text{CaLVA}}: *0.46$	$c_{\text{thresh}} = 0.968$
	$V_{\text{offma},\text{CaLVA}}: -4.3 \text{ mV}; V_{\text{sloma},\text{CaLVA}}: *1.14; V_{\text{offha},\text{CaLVA}}: -4.4 \text{ mV}; V_{\text{sloha},\text{CaLVA}}: *1.04;$ $\tau_{\text{m}^*,\text{CaLVA}}: *0.53; \tau_{\text{h}^*,\text{CaLVA}}: *0.46$	$c_{\text{thresh}} = 2.000$
<i>SCN1A</i> [14]	$V_{\text{offm},\text{Nat}}: -0.3 \text{ mV}; V_{\text{offh},\text{Nat}}: +5 \text{ mV}; V_{\text{slo}m,\text{Nat}}: *1.15; V_{\text{slo}h,\text{Nat}}: *1.23$	$c_{\text{thresh}} = 0.049$
<i>SCN1A</i> [102]	$V_{\text{offm},\text{Nat}}: +2.8 \text{ mV}; V_{\text{offh},\text{Nat}}: +9.6 \text{ mV}; V_{\text{slo}m,\text{Nat}}: *0.984; V_{\text{slo}h,\text{Nat}}: *1.042$	$c_{\text{thresh}} = 0.063$
<i>SCN1A</i> [105]	$V_{\text{offm},\text{Nat}}: -4.0 \text{ mV}; V_{\text{offh},\text{Nat}}: -5.8 \text{ mV}; V_{\text{slo}m,\text{Nat}}: *0.92; V_{\text{slo}h,\text{Nat}}: *1.13; \tau_{\text{h}^*,\text{Nat}}: *1.47$	$c_{\text{thresh}} = 0.273$
<i>SCN1A</i> [105]	$V_{\text{offm},\text{Nat}}: -8.1 \text{ mV}; V_{\text{offh},\text{Nat}}: +2.2 \text{ mV}; V_{\text{slo}m,\text{Nat}}: *0.97; V_{\text{slo}h,\text{Nat}}: *0.97; \tau_{\text{h}^*,\text{Nat}}: *1.59$	$c_{\text{thresh}} = 0.037$
<i>SCN1A</i> [13]	$V_{\text{offm},\text{Nat}}: +6.0 \text{ mV}; V_{\text{slo}m,\text{Nat}}: *1.16; \tau_{\text{h}^*,\text{Nat}}: *1.29$	$c_{\text{thresh}} = 0.129$
<i>SCN1A</i> [65]	$V_{\text{offm},\text{Nat}}: +10.0 \text{ mV}; V_{\text{offh},\text{Nat}}: -0.6 \text{ mV}; V_{\text{slo}m,\text{Nat}}: *1.15; V_{\text{slo}h,\text{Nat}}: *1.14$	$c_{\text{thresh}} = 0.062$
<i>HCN1</i> [44]	$V_{\text{offm}^*,\text{h}}: -26.5 \text{ mV}; V_{\text{slo}m^*,\text{h}}: *0.64$	$c_{\text{thresh}} = 0.296$ *
<i>HCN1</i> [53]	$V_{\text{offm}^*,\text{h}}: -25.9 \text{ mV}; V_{\text{slo}m^*,\text{h}}: *0.6$	$c_{\text{thresh}} = 0.282$ *
	$V_{\text{offm}^*,\text{h}}: +17.7 \text{ mV}; V_{\text{slo}m^*,\text{h}}: *0.6$	$c_{\text{thresh}} = 0.807$
<i>HCN1</i> [108]	$V_{\text{offm}^*,\text{h}}: +3.9 \text{ mV}; \tau_{\text{m}^*,\text{h}}: *0.88$	$c_{\text{thresh}} = 1.226$

Synaptic Alterations

ASSR deficits in schizophrenia have previously been attributed to changes in synaptic transmission, mainly at GABAergic synapses [32]. To compare the effects of the cell-intrinsic changes to excitability introduced by the genetic variants considered in this study to these alterations at the synaptic level, we implemented the two most commonly modelled synaptic alterations [104, 70]: 1) a reduction of GABA levels due to a decrease in the expression of GAD67 [32], a GABA precursor, operationalized as a 25% reduction of the maximal GABA conductance g_{max} at GABAergic synapses (referred to as the *Gmax* condition) and 2) an increase of the decay time constant τ_{decay} at GABAergic synapses, as a result of a decrease of GAT1 [32], an enzyme responsible for GABA reuptake, operationalized as an increase from 8 ms to 25 ms (referred to as the *IPSC* condition).

Data analysis

The simulated local field potential (LFP) was recorded using NetPyNE's LFP recording capabilities with a recording time step of 0.1 ms. From the simulated LFP signals we calculated two measures capturing the degree of synchronous network activity: 1) the power spectral density and 2) the inter-trial phase coherence of the entrained network oscillation.

To compute the power spectral density, we used Welch's periodogram method (using the implementation in `matplotlib.pyplot`). Specifically, the power spectral density P_{xx} of the signal is computed as

$$P_{xx} = |fft(x)|^2$$

where x is the input signal and fft the Fast-Fourier Transform.

For the inter-trial phase coherence, we band-pass filtered the simulated LFP signal around the frequency of interest, i.e. 40 Hz \pm 2.5 Hz, and computed an analytic signal using the Hilbert transform. The analytic signal $x_a(t)$ can be expressed in polar coordinates as

$$x_a(t) = x_m(t)e^{j\phi(t)}$$

where $x_m(t)$ is called the instantaneous amplitude or envelope and $\phi(t)$ is called the instantaneous phase. We calculated the instantaneous phase $\phi_k(t)$ for each trial k of each subject j and then, for each subject j , the inter-trial phase coherence was calculated as

$$ITC(j, t) = \left| \frac{1}{N} \sum_{k=1}^N e^{i\phi_k(t)} \right|$$

We then averaged the individual inter-trial phase coherences over time. The ITC is a normalized measure for which a value of 0 reflects maximal variability of phases across trials and a value of 1 no variability.

To further understand the effect the changes to individual parameters have on the gamma power, we correlated (Pearson correlation) the parameter changes (i.e. changes to *offm*, *offh*, *slo*, *sloh*, *taum* and *tauh*), for the model variants affecting the HVA Ca²⁺ channel with the change in evoked gamma power. Since we only included

relatively few model variants affecting the other channels (LVA Ca^{2+} , Nat and h), we omitted such an analysis for those channels.

Comparison of single model variants, combinations of model variants and synaptic alterations against the shared control was performed using Cumming estimation plots (see Figures 3, 4, and 5) and permutation t-tests (see Table 4) using the DABEST Python package [39]. 5000 bootstrap samples were taken; the confidence interval was bias-corrected and accelerated. The p values reported are the likelihoods of observing the mean differences, if the null hypothesis of zero difference is true. For each permutation p value, 5000 reshuffles of the control and test labels were performed. For more details see [39].

To compare our findings with our earlier findings [63], we calculated Pearson correlation coefficients between gamma power and delta resonance and PPI threshold, respectively. Specifically, delta resonance power from our earlier study was calculated as the median of the power spectrum amplitude at a baseline frequency of 1.5 Hz. The PPI threshold is the threshold synaptic conductance of 3000 simultaneously activated synapses for generating a second action potential if a suprathreshold stimulus was given to the same synapses 60 ms before. For further details see [63]. We have to note that delta resonance power in our earlier work was generated in a substantially different network model, although based on the same single cell model of layer 5 pyramidal cells, and that the PPI thresholds were generated with the full layer 5 pyramidal cell model from Hay et al. [37]. Importantly, the modelling of the SNP-like genetic variants followed exactly the same procedure as outlined here.

Supplementary Results

Single Cell Behaviour

The response of the two single cell models in response to somatic, and in the case of the layer 5 pyramidal cell model, also to dendritic input currents has been extensively studied in earlier studies [58, 103].

Network Behaviour

Noise-driven network To validate our control network model, we first performed simulations (20 subjects, 10 trials each subject, see Supplementary Section 4) with only the background noise as input and calculated the background firing rates for each population. The pyramidal cell population fired at an average rate of 7.81 Hz (standard deviation: 0.07 Hz) and the basket cell population at 6.18 Hz (standard deviation: 0.26 Hz), which is reasonably close to the range of 4.9 Hz (standard deviation: 0.5) reported for auditory cortex [42].

In order to explore the behaviour of the control network further, simulations with 20, 30 and 40 Hz drive were performed and the simulated local field potential was recorded. We explicitly went beyond purely driving the network with 40 Hz stimuli and included drive at 20 and 30 Hz, which are routinely used in experimental ASSR studies in patients with schizophrenia (e.g. [51, 104]). As can be seen in Supplementary Figure

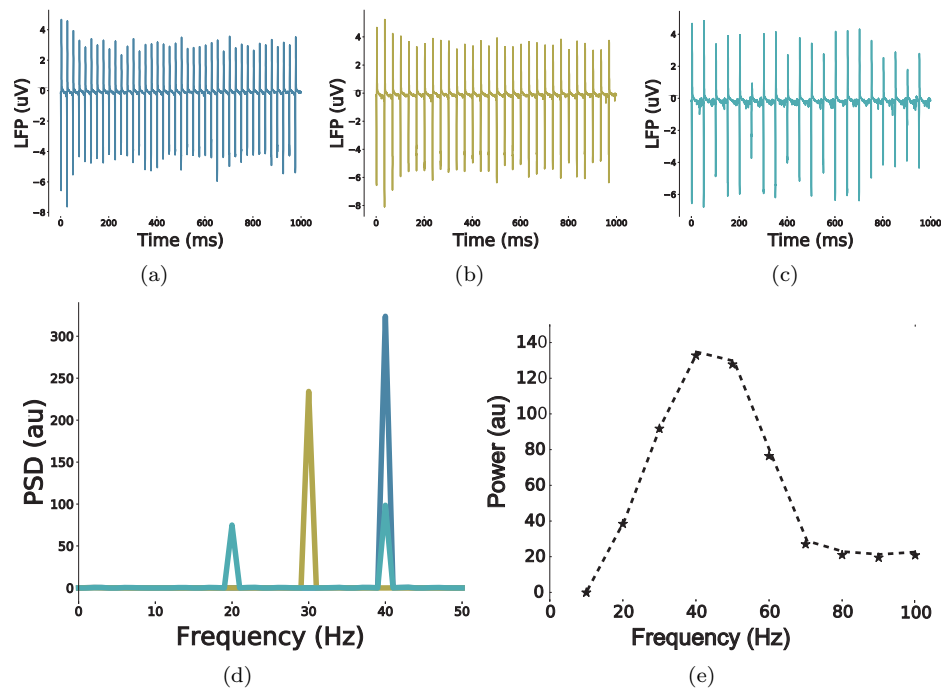


Figure 7: **Validation of the control network model.** (a) Simulated LFP in response to a 40 Hz drive. (b) Simulated LFP in response to a 30 Hz drive. (c) Simulated LFP in response to a 20 Hz drive. (d) Power spectral densities of the three signals from (a)-(c). (e) Peak power for simulations with 10-100 Hz drive to only the FS basket cell populations (Note that in the standard ASSR condition (a-d) both cell populations receive drive input).

7, the network clearly entrains to the driving frequency, the power at drive frequency is highest for 40 Hz and lowest for 20 Hz and for 20 Hz drive, a clear 40 Hz component is visible, thus, the network replicates experimental studies [51, 104] and previous modeling studies [104, 70, 68, 86].

Furthermore, Cardin et al. [11] show that rhythmic optogenetic drive of FS cells specifically enhances power in the gamma range. In our control network model, we rhythmically drive FS cells (with a sinusoidal current instead of optogenetic drive though) with frequencies between 10 and 100 Hz, in steps of 10 Hz, while excitatory cells only receive noise drive. Similarly to Cardin et al. [11], we see that periodically stimulating FS cells enhances LFP power especially in the gamma frequency range (30-60 Hz) (Supplementary Figure 7). This replicates experimental [11] and modelling findings [103] and further underpins the validity of the employed network model.

Single Variant Effects

Supplementary Table 4 shows the mean differences and permutation statistics for selected single model variants and model variant combinations as well as the two synaptic alterations.

	sample size	mean difference	99% CI interval	permutation p-value
Ca7	20	-29.5	[-30.9, -28.2]	< 0.001
HCN2	20	-21.9	[-23.7, -20.1]	< 0.001
Ca74	20	-16.7	[-18.1, -15.2]	< 0.01
C1	20	-49.7	[-51.0, -48.4]	< 0.001
C2	20	-61.9	[-62.2, -60.6]	< 0.001
C3	20	-67.2	[-68.5, -65.9]	< 0.001
Reduced g_{max}	20	-88.4	[-90.0, -86.7]	< 0.001
Increased τ_{inh}	20	-181.0	[-183.0, -170.0]	< 0.001

Table 4: **Permutation test statistics.** Comparison against a shared control with a sample size of $n = 20$. 5000 bootstrap samples were taken; the confidence interval is bias-corrected and accelerated. The p values reported are the likelihoods of observing the mean differences, if the null hypothesis of zero difference is true. For each permutation p value, 5000 reshuffles of the control and test labels were performed. For more details see [39].

Several experimental [51, 104] and modeling [104, 70, 68, 71] studies have found an increase in the subharmonic 20 Hz component in response to 40 Hz drive in SCZ. Therefore, we also quantified 20 Hz power in our exploration of the single model variant effects. As can be seen in Supplementary Figure 8, single model variants had hardly any effects on the 20 Hz component.

Lastly, we correlated parameter changes to gamma reduction for the model variants affecting I_{CaHVA} . As can be seen in Supplementary Figure S9 the change in evoked gamma power was positively correlated with the offset of the activation variable m but there was no significant correlation with other parameter changes.

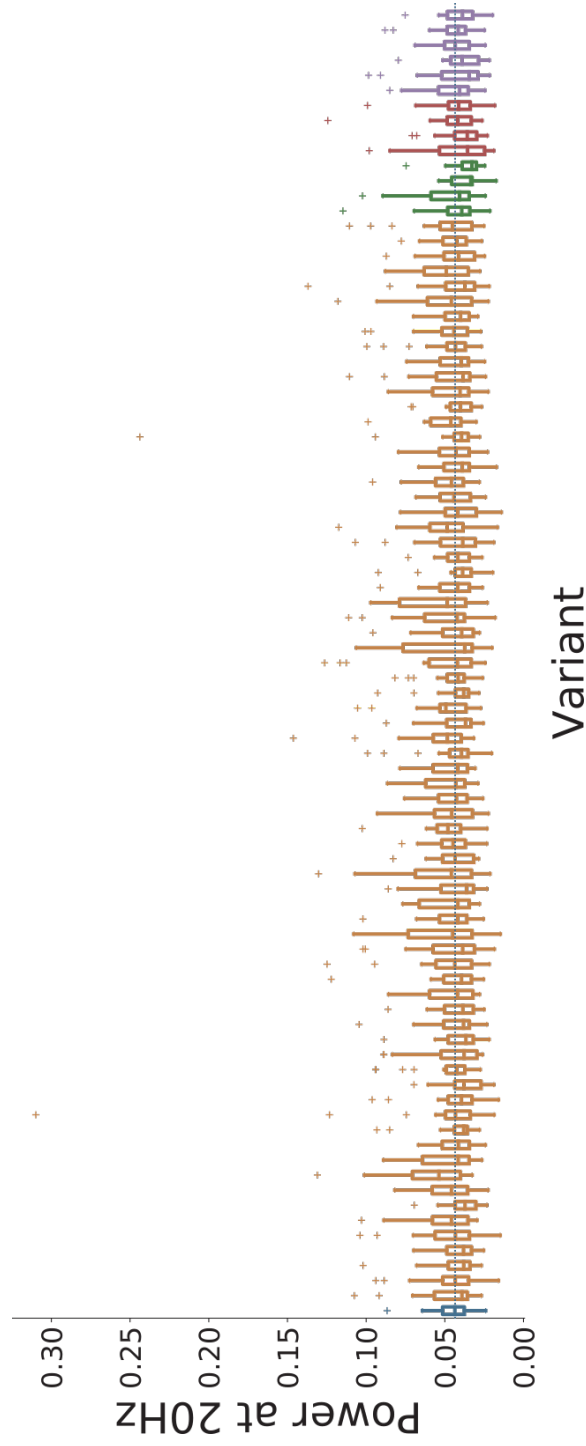


Figure 8: **Overview of the 2040 measure for all variants.** Control network in blue, Ca^{2+} channel variants affecting Ca_{HVA} in orange, affecting Ca_{LVA} in green, HCN variants in red, and SCN variants in purple. Solid lines represent the mean, box edges the 25 and 75 percentile, respectively, the whisker extend to 2 standard deviations and + depict outliers. The dashed blue line represents the control mean.

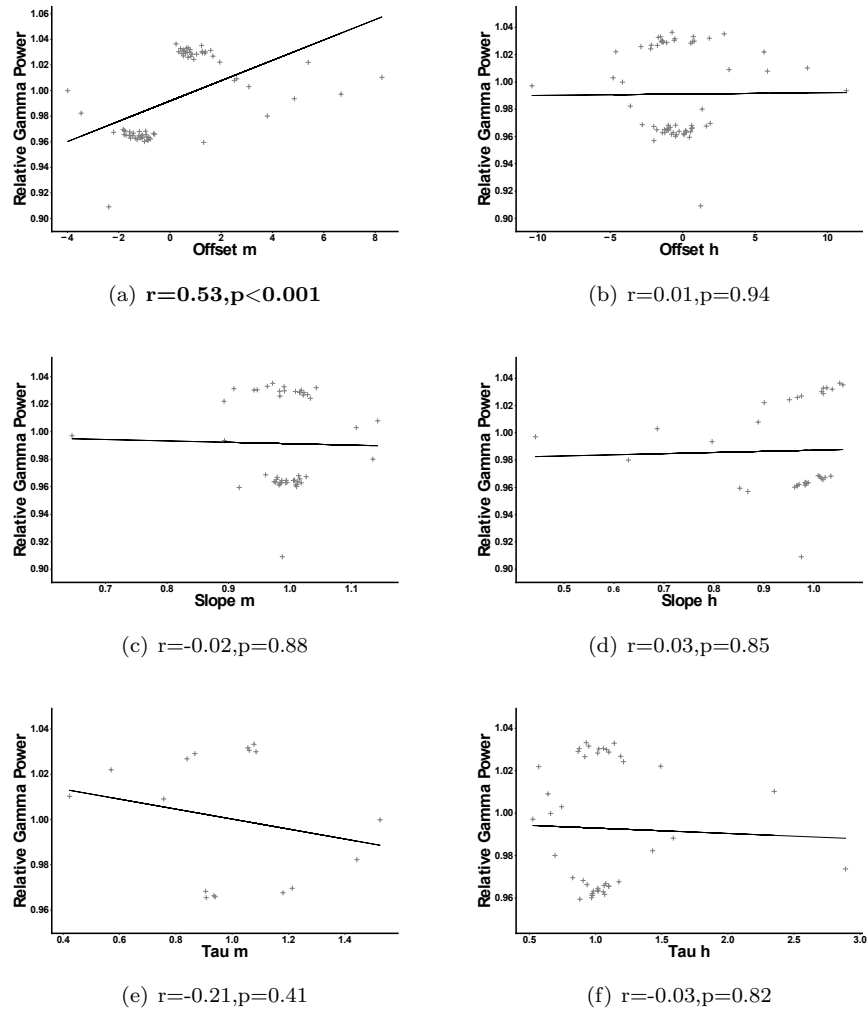


Figure 9: **Correlation of parameter changes with gamma reduction.** Pearson correlation between scaled changes of a parameter of the high-voltage activated Ca^{2+} channel and relative gamma power (i.e. the ratio of gamma power of a variant and the gamma power of the control).

Data Availability

All code to simulate the computational model, to analyze the data and to create the figures is available at <https://github.com/ChristophMetzner/Gene-Variants-and-Gamma>. The model code will also be made available through ModelDB (<https://senselab.med.yale.edu/modeldb/>) upon publication. Furthermore, the model will also be included in the ASSRUnit package, which is designed for automated validation and comparison of models of ASSR deficits in psychiatric disorders [69].

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