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Anesthetic action on extra-synaptic receptors: effects in neural population models of EEG activity.

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General anesthesia: from theory to experiments

2 ABSTRACT

1

3 The role of extra-synaptic receptors in the regulation of excitation and inhibition in the brain has attracted increasing attention. Because activity in the extra-synaptic receptors plays a role in 4 regulating the level of excitation and inhibition in the brain, they may be important in determining 5 the level of consciousness. This paper reviews briefly the literature on extra-synaptic GABA and NMDA receptors and their affinity to anesthetic drugs. We propose a neural population 6 7 model that illustrates how the effect of the anesthetic drug propofol on GABAergic extra-8 synaptic receptors results in changes in neural population activity and the electroencephalogram 9 (ÉEG). Our results show that increased tonic inhibition in inhibitory cortical neurons cause a 10 dramatic increase in the power of both δ and α bands. Conversely, the effects of increased tonic 11 inhibition in cortical excitatory neurons and thalamic relay neurons have the opposite effect and 12 decrease the power in these bands. The increased δ -activity is in accord with observed data 13 for deepening propofol anesthesia; but is absolutely dependent on the inclusion of extrasynaptic 14 (tonic) GABA action in the model. 15

16 Keywords: GABA receptor, neural mass, propofol, power spectrum, general anesthesia

1 INTRODUCTION

17 General anesthesia is used daily to enable surgery, but its underlying mechanisms of action are still largely a mystery. In recent decades there have been successful efforts to reveal the drug action on 18 single receptors (1, 2, 3), however their effect on neural populations, networks of neural populations, 19 and brain areas, still remains unsolved. To explain the underlying neural mechanism during the loss 20 of consciousness, two prominent hypotheses are the loss of integration information, developed by 21 Tononi (4, 5, 6), and a sharp phase transition of the brain activity involving a drop of neural activity, 22 put forward by Steyn-Ross et al. (7, 8). These hypotheses are not mutually exclusive. For instance, 23 a recent experimental study on the effects of propofol on neural activity measured at various spatial 24 scales (9) has revealed both decreased functional connectivity between brain areas and a dramatic 25 drop of neuron firing rates after loss of consciousness. A large amount of experimental literature 26 has revealed characteristic spectral signal changes in electroencephalographic data (EEG) and Local 27

Field Potentials (LFPs) during general anesthesia (10, 11, 9, 12, 13). Moreover, several previous 28 theoretical studies have proposed neural models to explain certain EEG signal features observed during 29 anesthesia (14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25). Although these studies may incorporate realistic 30 neurobiological details of the brains' network topology and neuronal function, they have simplified 31 dramatically the anesthetic action by considering only synaptic excitatory and inhibitory receptors. There 32 is a growing amount of experimental research that has revealed the importance of extra-synaptic receptors 33 (ESR) for neural interactions in general (3, 26), and for anesthesia especially, see (27, 2) and references 34 35 therein. To elucidate the role of ESR in the context of anesthesia, one approach might be to do a theoretical study 36 37 of a realistic neural population model which reproduces the characteristic signal features observed in EEG. 38 To perform such a theoretical study, it is necessary to incorporate physiological properties of extrasynaptic receptors into neural population models. 39 40 Gamma-aminobutyric acid (GABA) receptors are a large and important class of ionotropic receptors. 41 These receptors are located in the neuron's membrane and respond to the neurotransmitter GABA by 42 43 opening Cl⁻ channels and inducing an inward hyperpolarising membrane current. This response may either be: phasic at synaptic receptors or, tonic at ESR which lie distant from synaptic terminals (28, 29, 44 30, 31, 32, 33). The phasic response evolves on a time scale of 10-200 ms whereas tonic response evolves 45 on a much longer time scale (34, 35). 46

The precise biochemical origin of tonic inhibition is still heavily debated (36, 37). A rather simple and intuitive model explains the tonic current as a spillover of excess neurotransmitter from synapses. This is due to incomplete GABA uptake by nearby synaptic GABA_A-receptors. The remaining neurotransmitter is thus able to diffuse to more distant GABA_A-receptors via extracellular space (36, 31, 38, 39). This spillover may explain the longer time scale of tonic responses found experimentally. In addition, this explanation implies that even small concentrations of neurotransmitters are sufficient to generate tonic activity because of the high sensitivity of ESRs.

The effect of extra-synaptic receptors on the dendritic activity has not attracted much attention. This may because there are only a relatively small number of such receptors as compared to synaptic receptors (40, 36). Moreover, only recently have experimental studies been able to classify and localize different subtypes of GABA_A receptors (31, 36). GABA_A receptors are pentameric ligand-gated ion channels and it has been found that δ -sub units of GABA_A receptors occur exclusively at ESRs (32, 36, 41, 42, 43). This indicates a specific role of these receptors for the neural information processing in general with specific implications in diseases (3) and consciousness (40).

Tonic inhibition induced by extra-synaptic $GABA_A$ -receptors represents a persistent increase in the 61 cell membrane's conductance. On the single neuron level, this diminishes the membrane time constant 62 and, consequently, reduces the size and duration of excitatory post-synaptic potentials propagating on 63 the dendrite. Hence tonic inhibition reduces the excitability of the membrane and increases the effective 64 65 firing threshold (36). At the neural population level, extra-synaptic receptors affect the excitability of 66 interneuron-pyramidal cell networks and thus modify network oscillations (30). Kopanitsa (40) argues that the sustained spatially widespread tonic inhibition is energetically more effective for the system to 67 68 diminish neural population activity than short-lasting local phasic inhibition, since lower neurotransmitter concentrations are sufficient. The critical factor in this mechanism is the the relatively high sensitivity of 69 ESRs to modulations by anesthetic agents (36, 44, 28, 45). The brain areas that have been shown to be 70 affected by anesthetic-induced tonic inhibition are the hippocampus (46), brain stem (47), cerebellum (45) 71 and the thalamus (32). Since these areas are supposed to play a role in general anesthesia (2), extra-72 73 synaptic receptors may mediate clinical anesthetic effects, such as hypnosis and amnesia (48). Thus it is reasonable to argue that GABA_A ESRs set the background inhibition of neural populations and the brain 74 network and mediate slow consciousness phenomena, such as loss of consciousness, sleep or arousal (40). 75 76

77 Converse to GABAergic receptors, N-methyl-D-aspartate (NMDA) receptors respond to the 78 neurotransmitter glutamate by excitatory inward Na⁺ and Ca²⁺ currents and K⁺ outward currents. The

79 response of NMDA receptors to glutamate depends on their spatial location with respect to synaptic terminals and the presence of co-agonists. A recent experimental study has revealed that the population 80 of NMDA receptors, which are close to synaptic terminals, are primarily activated by the co-agonist d-81 82 serine in the presence of glutamate; while extra-synaptic NMDA receptors (more distant from the synaptic 83 terminals) respond primarily to glutamate and the co-agonist glycine (49, 50). D-serine and glycine are endogenous amino acids found naturally in the brain (d-serine is a derivative of glycine). Similar to 84 GABAergic ESRs, it has been shown that there exists a significant ambient glutamate concentration which 85 induces a tonic excitatory current (51, 52). This current is evoked primarily at extra-synaptic NMDA 86 87 receptors (53) and may be regulated by other cells, such as neighbouring astrocytes (52, 54) which control 88 glutamate uptake and also synthesize d-serine (55).

Commonly-used GABAergic anesthetic drugs directly modify the corresponding receptors. However 89 90 various anesthetics are also known to affect the endogeneous co-agonists of NMDA receptors (56, 49, 57). Hence, the possible anesthetic effect on NMDA receptors is more complex and indirect than for 91 92 GABAergic ESRs. There is a large class of NMDA receptor antagonists, that inhibit the excitatory action of NMDA receptors. These anesthetics induce so-called dissociative anesthesia (58) leading to amnesia 93 94 and analgesia without depressing respiration, but also characterised by distorted perceptions of sight and sound and feelings of dissociation from the environment. An example of a dissociative anesthetic drug 95 is the inhalational anesthetic xenon which - amongst other actions - binds primarily to the extra-synaptic 96 97 glycine site of NMDA receptors (59) and attenuates long term potentiation present in the hippocampus by 98 reducing extrasynaptic receptor currents (60). 99

To understand how the anesthetic effect of extra-synaptic receptor activity on the microscopic single 100 101 neuron scale could lead to changes in EEG and behavior that can be observed at macroscopic scales, it is necessary to establish a bridge between the two scales. This bridge may be formulated as a dynamical 102 theoretical model. Neural population models represent a good candidate for a dynamic description of 103 neural activity at an intermediate mesoscopic scale (61, 62, 63). These models describe properties of 104 105 ensembles of neurons, such as the mean firing rate and the mean dendritic current (64), whilst their output variables can be strongly linked to macroscopic experimental quantities such as Local Field Potentials 106 (LFPs) and EEG (63, 65). An increasing number of theoretical studies have used neural population 107 models to describe signal features in LFPs and EEG observed during anesthesia (21, 17). Most of these 108 109 studies take into account anesthetic action on excitatory and/or inhibitory synapses (14, 66, 7, 67, 68) 110 while few consider extra-synaptic receptors (69). This link between the synaptic receptor properties in an ensemble of neurons and the average population dynamics is straight-forward, since classical neural 111 population models already involve the average synaptic response function. The situation is different for 112 extra-synaptic receptors, since their action is not incorporated into the classical models. A very recent 113 work has filled this gap (24). This theoretical work demonstrated a method to include mathematically 114 115 extra-synaptic GABA_A receptor action in neural population models; which enables researchers to study how changing anesthetic extra-synaptic receptor action modifies spectral features in the EEG, which might 116 then be observed experimentally. 117

118

The current work uses a cortico-thalamic neural population model involving anesthetic synaptic inhibition with a well-established connection topology; and then extends this model by including the effects of extra-synaptic GABAergic receptor action in the presence of the anesthetic drug propofol. With the help of this model, we demonstrate the role of extra-synaptic GABAergic inhibition, and the importance of tonic inhibition in the cortical inhibitory neuronal population, in explaining experimental EEG power spectra.

2 MATERIAL & METHODS

EEG DATA

We re-analysed previously-obtained experimental data from subjects that had been given a short propofol 125 126 anesthetic. The details of the methods can be found in (70). In brief, after obtaining regional ethical committee approval and written informed consent, five healthy subjects (mean age 27.7 yrs, four males) 127 128 were studied. They were on no psychoactive drugs and had been starved for at least 6 hours prior to 129 the study. They were monitored and managed as per clinical anesthesia, according to the Australia and 130 New Zealand College of Anaesthetists best practice guidelines. The induction consisted of an intravenous infusion of propofol at 1500 mg/hr until the subject no longer responded to verbal command. Typically 131 this occurred about 5 minutes into the infusion. The estimated effect-site concentrations of propofol were 132 calculated using standard population-based pharmacokinetic models. 133

The EEG was acquired using the Electrical Geodesics 128 channel Ag/AgCl electrode system (Eugene, 134 135 CO, USA) referenced to Cz. Electrode impedances were below 30 KOhm (100 MOhm input impedance 136 amplifier). The sampling frequency was 250 Hz, with a 0.1 - 100 Hz analogue band pass filter, and A-D conversion was at 12 bits precision. The EEG data were re-referenced to a grand mean, and band-pass 137 filtered using 3-rd order Butterworth filters 0.2 - 45 Hz to eliminate line-noise. An additional Whittaker 138 filter was applied to reduce movement and blink artifacts. The power in each frequency was obtained 139 applying a short-time Fourier transform with a moving window of 60 sec and 54 sec overlap. The power 140 141 spectra have been computed 1 min before infusion start (t = 1 min) and 4 min after infusion (t = 5min). For visualization reasons, these power spectra at different time instances have been smoothed by a 142 running average over frequencies with a 1 Hz window and a 0.017 Hz frequency step. 143

THALAMO-CORTICAL MODEL

144 The body of the model (71, 72) is based on a population-level description of a single thalamo-cortical module comprising four populations of neurons, namely excitatory (E) and inhibitory (I) cortical 145 population, a population built of thalamo-cortical relay neurons (S) and of thalamic reticular neurons 146 (R). The details of the model and the nominal parameter values are taken from a previous work (72, 73). 147 148 This model is based on the original idea of Lopes da Silva et al., stating that the α -rhythm represents 149 the noisy thalamic input signal band-pass filtered by feedback-connected cortical and thalamic neural populations (74). Here we just briefly describe the key concepts of the model. The average soma 150 membrane potential denoted by V_a , for a = E, I, S, R is modeled by 151

$$V_a(t) = \sum_{b=E,I,R,S} h_b(t) \otimes \nu_{ab} \phi_b(t - \tau_{ab}), \tag{1}$$

where \otimes represents the temporal convolution, $h_b(t) = H_b \bar{h}_b(t)$ where $\bar{h}_b(t)$ denotes the mean synaptic response function defined by

$$\bar{h}_b(t) = \frac{\alpha \beta_b}{\alpha - \beta_b} \left(e^{-\beta_b t} - e^{-\alpha t} \right), \tag{2}$$

and α and β_b (with units Hz) are the synaptic rise and decay rates of the synaptic response, respectively. The synaptic decay rates and synaptic response functions depend on the source neurons of type *b* only and are independent of the target neurons. The constant prefactor H_b defines the response function amplitude subject to the anesthetic concentration. Here, we assume identical excitatory synaptic receptors with constant rise and decay rate. Inhibitory synaptic receptors are also assumed to exhibit identical constant rise and decay rates while their decay rates depend on the anesthetic concentration. This strong approximation is taken from a previous study (68) to be able to compare our results, while preliminary 161 studies with more realistic parameters show similar results (not shown).

162 The constants ν_{ab} are the strengths of the connections from population of type *b* to population of type *a* 163 (in mVs), and ϕ_b is the average firing rate of the population *b* (in Hz). The connections between cortex 164 and thalamus are associated with a same nonzero time delay, $\tau_{ab} = \tau$, while the delay term is assumed to 165 be zero within the cortex and within the thalamus (75).

By virtue of long-range axonal projections of excitatory cortical neurons and by assuming the spatiallyhomogeneous dynamics on the cortex, the average firing rate ϕ_E obeys the damped oscillator equation

$$D\phi_E = S(V_E),\tag{3}$$

168 where the operator D is defined as

$$D = \left(\frac{1}{\gamma}\frac{\partial}{\partial t} + 1\right)^2 \tag{4}$$

169 and γ is the cortical damping rate. It is assumed that the spatial spread of activity is very fast in other 170 populations and the activity variable can be approximated by a sigmoidal function as $\phi_b = S(V_b)$, for 171 b = I, S, R. Conversely to the original model (72, 73, 75) we use a more realistic transfer function 172 derived from properties of type I-neurons given by (24)

$$S(V_a) = Sig(V_a, 0) - Sig(V_a, \rho),$$
(5)

173 with

$$Sig(V_a,\rho) = \frac{S_a^{max}}{2} \left(1 + \operatorname{erf}\left(\frac{V_a - \theta_a - \rho\sigma^2}{\sqrt{2}\sigma}\right) \right) e^{-\rho(V_a - \theta_a) + \rho^2 \sigma^2/2},\tag{6}$$

with $\sigma = 10 \text{ mV}$ and $\rho = 0.08 \text{ mV}^{-1}$, where σ is related to the standard deviation of firing thresholds, the parameter $\rho < \infty$ reflects the properties of type I-neurons, S^{max} is the maximum population firing rate, and θ_a is the mean firing threshold of neurons in population a. In contrast to the standard transfer function given in (72), the transfer function in Eq. (6) is not anti-symmetric to its inflection point anymore (27) and exhibits a larger nonlinear gain (slope) for large potentials $V_a > \theta_a$ compared to small potentials $V_a < \theta_a$. This asymmetry results from the firing properties of type-I neurons, see (24, 27) for more details. For $\rho \to \infty$, the sigmoid function becomes the conventional symmetric transfer function.

181 The external input to the system is considered as a non-specific input to thalamo-cortical relay neurons 182 as

$$\phi_N = \langle \phi_N \rangle + \sqrt{2\kappa\xi(t)},\tag{7}$$

183 where $\langle \phi_N \rangle$ indicates its mean value and $\xi(t)$ is a zero average Gaussian white noise and κ is the noise 184 intensity.

The power spectrum characterises small fluctuations about the resting state of the system defined by $dV_a(t)/dt = 0$. Following (63, 72), it is assumed that the activity of excitatory cortical neurons generates the EEG, and due to the specific choice of external input to thalamo-cortical relay neurons, the power spectrum of the EEG is related to the Greens function of linear deviations about the resting state by (25)

$$P_E(\omega) = 2\kappa \sqrt{2\pi} \left| \tilde{G}_{1,3}(\omega) \right|^2, \tag{8}$$

189 in which $P_E(\omega)$ depends just on one matrix component of the matrix Greens function $G(\omega)$, see the 190 Appendix for its definition. We point out that the subsequent power spectrum analysis is based on 191 Eq. (8) and changing a system parameter, such as the factor p, changes the resting state, the corresponding 192 nonlinear gains and consequently the power spectrum. In addition, the power spectrum analysis is valid 193 only if the resting state is stable and hence the fluctuations do not diverge. We have taken care of this 194 additional condition and all given parameters guarantee the existence and stability of the resting state.



Figure 1. The schematic of a thalamo-cortical module. The blue arrows indicate excitatory connections while the red connections with filled circle ends denote inhibitory connections. The symbols E, I, S, and R denote the excitatory and inhibitory cortical neurons, thalamo-cortical relay, and thalamic reticular neurons respectively. In addition thalamocortical and corticothalamic connections exhibit the same nonzero time delay τ .

EFFECT OF PROPOFOL ON NEURAL POPULATIONS

195 In order to mimic anesthetic action, we consider the general anesthetic propofol which affects synaptic 196 and extra-synaptic GABAergic receptors. We assume that the decay rate of inhibitory synapses is identical 197 in all neural populations under study, and decreases with increasing propofol concentration in accordance 198 with experimental findings (76, 67). Mathematically, such a dependence on the anesthetic concentration 199 can be taken into account by a concentration factor $p \ge 1$ and $\beta_b = \beta_0/p$ while increasing p reflects an 200 increase of the on-site concentration of propofol (21, 17, 68).

Since propofol has been shown to retain the amplitude of inhibitory synaptic response functions (76), one can define $H_b = \Gamma(\alpha, \beta_0) / \Gamma(\alpha, \beta_b)$ for b = I, R, where

$$\Gamma(\alpha,\beta) = \frac{\alpha\beta}{\alpha-\beta} \left[(\alpha/\beta)^{\frac{-\beta}{\alpha-\beta}} - (\alpha/\beta)^{\frac{-\alpha}{\alpha-\beta}} \right],$$

i.e., $\Gamma(\alpha, \beta_b) = \bar{h}_b(t_0)$ is the peak amplitude of $\bar{h}_b(t)$ at time $t_0 = \ln(\alpha/\beta_b)/(\alpha - \beta_b)$. Thereby the maximum height of $h_b(t)$ is $h_{max} = \Gamma(\alpha, \beta_0)$ which is independent of the action of propofol (68). Moreover, since it is assumed that propofol does not act on excitatory synaptic transmission, $H_b = 1$ and $h_b(t) = \bar{h}_b(t)$ for b = E, S.

The GABAergic ESR tonic inhibition can be represented in the model as a constant shift of the firing threshold in neural population models (24). For simplicity, we assume a linear relationship between the anesthetic concentration parameter p and the extra-synaptic threshold shift

$$\theta_a = \theta_0 + (p-1)k_a \tag{9}$$

210 with the unique firing threshold $\theta_0 = 15 \text{ mV}$ identical for all populations in the absence of propofol 211 and the extra-synaptic anesthetic sensitivity $k_a > 0$. Here, $(p-1)k_a$ is the tonic inhibition induced by 212 extra-synaptic action which depends linearly on the propofol concentration. Future experimental studies

213 may motivate a more realistic relationship of threshold shift and the anesthetic concentration parameter.
214 Summarizing, synaptic and extra-synaptic inhibition, and hence anesthetic action, is present in the cortical

215 populations E and I and in the thalamic population of relay neurons S.

POWER SPECTRUM

The present study examines the effect of tonic inhibition in various populations E, I, S on the power spectrum of neural activity in cortical excitatory neurons, i.e., population E. We will focus on the power in the δ - and α - frequency ranges in the interval [0.5Hz-4Hz] and [8Hz-12Hz], respectively.

219

220 The subsequent analysis reveals power peaks in these frequency ranges, whose magnitude changes with 221 the level of tonic inhibition. These power peaks exhibit a maximum of power, expressed mathematically as a local maximum of the function $P_E(f)$ where P_E is taken from Eq. (8). The local maximum at frequency f_0 is defined as $dP_E/df = 0$, $d^2P_E/df^2 < 0$ computed at f_0 . If there is a local maximum of power in the 222 223 δ – frequency range, then δ – activity is present, whereas a missing local maximum in the δ – frequency 224 225 range indicates missing δ -activity. Since the magnitude and frequency of power peaks change with the propofol concentration and extra-synaptic threshold, the concentration factor p and the extra-synaptic 226 anesthetic sensitivity k_a are the parameters of the power spectrum, i.e. $P_E = P_E(p, k_a, f)$. 227 228

229 To illustrate the usefulness of this parametrization, let us assume a factor k_{a0} for which no δ -power peak exists in the power spectrum $P_E(p, k_{a0}, f)$, and k_{a1} , $k_{a1} > k_{a0}$ is the extra-synaptic anesthetic sensitivity leading to a spectral δ -power peak in $P_E(p, k_{a1}, f)$ with $dP_E(p, k_{a1}, f_{max})/df_{max} = 0$ where f_{max} is a frequency in the δ -frequency range. Mathematically, then the continuity of all model functions 230 231 232 233 and variables guarantee that there is a threshold for the emergence of δ -activity at a certain extrasynaptic anesthetic sensitivity $k_{a,thr}$ with $k_{a0} \leq k_{a,thr} \leq k_{a1}$. Consequently, if a threshold extra-synaptic 234 anesthetic sensitivity for δ -activity exists, then the variation of model variables about this critical point 235 guarantees the emergence of δ -activity. This mathematical reasoning allows us to investigate conditions 236 under which δ -activity may emerge. 237

3 **RESULTS**

Figure 2 illustrates how the EEG power spectrum depends on the concentration of propofol for a single subject. After starting the infusion at t = 0 min, the estimated propofol effect-site concentration increases gradually with time (Fig. 2(A)); resulting in increased power in the δ - and α -frequency ranges (Fig. 2(B)). Over the period of the spectrogram the subject has become progressively more sedated; until a t = 5 min the subject no longer responds to verbal command but would still be responsive to nociceptive stimuli. Figure 2(C) shows the power spectra in the awake and sedation conditions. We observe a power enhancement primarily in the δ - and α -frequency ranges.

245

To understand how propofol might enhance δ – and α – power, we study the power spectrum of our 246 247 theoretical model for different anesthetic concentration levels and examine the impact of adding tonic 248 inhibition via extra-synaptic GABA_A receptors. Figure 3(A) shows the interaction between propofol and tonic inhibition in the cortical inhibitory neuronal population. If we set the tonic inhibition to zero $(k_I =$ 249 0mV), we observe a decrease in spectral power as propofol concentrations increase (i.e. the power moves 250 251 from the black line to the blue line in the figure). If we set the tonic inhibition to $(p-1) \cdot 15 \text{ mV}$ we see the opposite effect - there is an increase of δ - and α -power (black line to red line), with increasing 252 253 propofol concentration.



Figure 2. Electroencephalographic data observed under anesthesia sedation in a single subject while increasing the propofol concentration. (A) Blood plasma concentration of propofol with respect to administration time. (B) Spectrogram of frontal EEG power. The vertical lines denote time windows well before the administration (left line) and at about 5min after the start of propofol infusion (right side); (C) Power spectra computed before the infusion of propofol (black) and 5 min after the start of infusion (red).

Previous studies have indicated that extra-synaptic inhibition in thalamic relay neurons may control the level of inhibition in the brain (3). However Fig. 3(B) reveals that adding a non-zero tonic inhibition in the thalamic relay neurons causes a decrease in the spectral power, similar to the previous case of absent tonic inhibition in the inhibitory cortical neurons.

It is well-known that GABAergic anesthetics change the EEG from high frequency-low amplitude signals to low frequency-high amplitude signals (77, 78). Figure 3(C,D) show simulated time series in the absence and presence of tonic inhibition in cortical inhibitory cells reproducing this experimental finding.

Our results elucidates that tonic inhibition in cortical interneurons and thalamic relay neurons affect 263 264 the cortical power spectrum differently. This finding is similar to results of a previous computational neural population study of a cortico-thalamic feedback single-neuron model (69). Figure 4 shows how 265 the resting membrane potential (Fig. 4(A)) and the nonlinear gain (Fig. 4(B)) in the cortical excitatory 266 population change with differing extra-synaptic anesthetic sensitivity in cortical inhibitory neurons (k_I) 267 and in the thalamic relay neurons (k_S) . We observe that both the resting potential and the nonlinear gain 268 269 of cortical excitatory neurons increase when the cortical inhibitory extra-synaptic anesthetic sensitivity k_I increases, whereas resting potential and nonlinear gain of cortical excitatory neurons decrease when 270 271 the extra-synaptic anesthetic sensitivity in thalamic relay neuron k_S increases. Since the nonlinear gain is proportional to the systems responsiveness to external stimuli, the power enhancement in population I272 may be explained by the augmented responsiveness of the cortical excitatory neurons. This responsiveness 273 274 depends on the sub-circuit in which the neurons are involved. Since relay neurons are part of the thalamo-275 cortical feedback loop, while cortical inhibitory neurons contribute to the cortical loop, the cell types respond differently to the thalamic input. Essentially assuming tonic inhibition in the population of 276 cortical excitatory neurons E, the study reveals a similar propofol concentration dependence of the power 277 spectrum, the resting state potential and the nonlinear gain as for the thalamic tonic inhibition S. This 278 279 shows the unique tonic inhibition effect in the cortical inhibitory neurons. 280

Figure 3(A) shows the power spectrum for single values of the extra-synaptic sensitivity k_I , for single values of the concentration factor p and fixed strength of self cortical inhibition ν_{ii} , while Fig. 4 gives more details on the role of extra-synaptic sensitivity for fixed values of the concentration factor p and fixed cortical self-inhibition. To understand better the interplay between tonic inhibition, synaptic inhibition



Figure 3. The theoretical EEG power spectrum in the baseline and in the sedation condition with and without tonic inhibition in the cortical inhibitory neurons I in (A) and the thalamic relay neurons S in (B) and corresponding simulated EEG time-series. In (A) the administration of propofol without tonic inhibition (blue line) attenuates the power spectrum compared to the baseline condition (black line) while the tonic inhibition (red line) increases the global power and generates oscillatory activity in the δ -frequency range. In (B) increasing the anesthetic concentration yields a global power decrease in the sedation condition without tonic inhibition (red line). In (A) and (B), the black lines indicate the EEG-spectral power in the baseline condition (p = 1), and the blue and red lines show the power spectrum in anesthesia condition (p = 1.125) in the absence ($k_a = 0$) and in the presence ($k_a = 15 \text{ mV}$) of tonic inhibition, respectively. (C) The simulated EEG time-series ($\phi_E(t)$ defined in Eq. (3)) in the absence of extra-synaptic effects, i.e. $k_E = k_I = k_S = 0 \text{ mV}$. (D) The EEG time-series in the presence of extra-synaptic action in cortical inhibition changes the EEGs from low-amplitude, high-frequency pattern to high-amplitude, low-frequency pattern. In addition, the strenght of self cortical inhibition is $\nu_{ii} = -1.8 \text{ mVs}$.

285 and the strength of cortical self-inhibition, Figure 5 shows the parameter pairs of synaptic inhibition p286 and the threshold of extra-synaptic sensitivity $k_{I,thr}$ at different self-inhibition levels, for which a peak in the δ -frequency range emerges. Recall that the $k_{I,thr}$ is the critical (smallest) value of extra-synaptic 287 sensitivity in cortical inhibitory neurons k_I , that lead to $dP_E/df = 0$, $d^2P_E/df^2 < 0$ computed at $f_{max} \in$ 288 δ -range, cf. the subsection on the power spectrum in section 2. Parameter values beyond the respective 289 290 curves lead to δ -activity power peaks. We observe that δ -activity always emerges for sufficiently strong tonic inhibition (large extra-synaptic sensitivity k_I) and sufficiently strong self-inhibition ν_{ii} , while the 291 weaker the self inhibition is the larger is the necessary extra-synaptic sensitivity or the synaptic inhibition 292 293 to generate δ -activity. Even for vanishing self cortical inhibition ($\nu_{ii} = 0$), mathematical analysis (not shown) reveals that there is still a δ -peak in the power spectrum for large enough synaptic or tonic 294 295 inhibition (for p or k_I large enough).

Moreover, Fig. 5 reveals a minimum tonic inhibition level (minimum value of k_I) beneath which no δ -power peak emerges, irrespective of the level of synaptic inhibition (*p*). This result indicates a major role of tonic inhibition in the generation of δ -activity, since it may support δ -activity even if the synaptic inhibition level is not sufficient to support it.

300



Figure 4. Increasing the tonic inhibition (factor k_a for a = I and S) affects the resting state of excitatory cortical neurons V_E^* (A) and the corresponding nonlinear cortical gain function (B). Here the anesthetic concentration factor is identical in the populations a = E, I and S to (p = 1.125). In addition, the strength of self cortical inhibition is $\nu_{ii} = -1.8$ mVs.

4 DISCUSSION

301 In the sedation phase, for modest concentrations of propofol, the EEG power spectrum exhibits an increase in the δ - and α -frequency ranges (Fig. 2) as found experimentally in the induction phase 302 of propofol anesthesia (79). One possible explanation for these phenomena is by postulating stronger 303 GABAergic potentiation within cortical inhibitory neurons than within cortical pyramidal neurons (68). 304 We hypothesize, that cortical GABAergic self inhibition plays a decisive role. Figure 3 reveals that 305 the power surge in these frequency ranges might also result from extra-synaptic tonic inhibition active 306 in cortical inhibitory neurons. Tonic inhibition increases the firing threshold and hence diminishes the 307 308 output of inhibitory neurons to excitatory cortical neurons, which then allows increased excitation in the excitatory population and a power surge in the EEG. Conversely tonic inhibition in the thalamic 309 relay cells does not induce power surge in EEG since augmented inhibition in the thalamic relay cells 310 yields diminished excitation in cortical excitatory neurons, leading to a decrease in EEG power. This 311 interpretation is corroborated by Fig. 4 which demonstrates augmented and diminished nonlinear gain 312 in cortical excitatory neurons assuming tonic inhibition in inhibitory and thalamic relay population, 313 314 respectively. This reflects enhanced and weakened response to the noisy thalamic external input, see previous theoretical studies (68, 25) for a similar line of argument. 315

Figure 3 clearly reveals the emergence of δ -activity caused by extra-synaptic tonic inhibition which is affirmed by the existence of a minimum level of extra-synaptic inhibition shown in Fig. 5. Conversely, α -activity appears to be much less sensitive to tonic inhibition since it is present for all tonic inhibition levels. One interpretation may be the generation of α -activity by the cortico-thalamic feedback as hypothesized theoretically (80) while δ -activity results from the cortical interaction of excitatory and inhibitory neurons. The exact origins of propofol-induced α - and δ -activity are not known for certain. We find that the α oscillations arise from thalamocortical resonances. These oscillations are



Figure 5. Parameter space for δ -power peak. The lines give the smallest (threshold) value of the extra-synaptic sensitivity $k_{I,thr}$ that induces δ - oscillations in the sedation condition with respect to the concentration factor p for different values of self-inhibitory connections ν_{ii} . The weaker the self cortical inhibition (the smaller $|\nu_{ii}|$), the higher the necessary level of propofol concentration (larger p) and the tonic inhibition (larger $(p-1) \cdot k_I$) to induce δ -activity.

323 commonly synchronous across widespread cortical regions and are not easily generated in isolated cortical 324 tissue (81, 82). This affirms the original model of Lopes da Silva et al. (74). However, our model results 325 are equivalent to results of other models describing α -activity by purely cortical interactions. We are not aware of a methodology ruling out one or the other model and this is not the aim of the present 326 work. Our work just reveals the additional possibility that the thalamus serves as a possible (indirect) 327 source of α -activity. Similarly, the origin of δ -activity is not clear, but slow activity does increase 328 329 at higher concentrations of propofol - which may be associated with decreasing α -waves as observed during desflurance general anesthesia (83). This is in keeping with δ -waves becoming more pronounced 330 331 as the cortico-thalamic systems becomes increasingly hyperpolarized. However there is a lot of variability between patients as regards the relative power of α - and δ -activity during general anesthesia; which 332 would suggest that the true explanation is more complex, and requires recognition of other factors such as 333 the one presented in this paper - the influence of the propofol on extra-synaptic inhibition. 334

335 Although anesthetic action on synaptic and extra-synaptic GABAergic receptors is different, both 336 actions diminish neural activity and hence increase inhibition. Figure 5 elucidates that strong enough extra-synaptic or synaptic inhibition induce δ -activity. Hence, one may argue that the level of inhibition 337 plays an important role while its origin, i.e. synaptic or extra-synaptic, plays a secondary role. This 338 interpretation corroborates the idea of the balance of excitation and inhibition as the major mechanism 339 in general anesthesia. This interpretation is in good accordance to previous experimental findings on the 340 important role of the balance of excitation and inhibition in brain network under anesthesia (84, 85). 341 342 Such global concepts as excitation-inhibition balance are attractive to describe complex processes in general anesthesia. For instance, anesthetics alter arousal in several pathways, such as the cholinergic 343 pathway (86) and the orexinergic pathway which has been identified to activate a complex functional 344 network controlling, inter alia, the emergence from unconsciousness (87). 345 346

Our theoretical study considers the anesthetic propofol and its corresponding action at synaptic and extra-synaptic GABAergic receptors only, whereas it is known that propofol induces inhibition at various other receptors as well (2, 88) including minor effects on NMDA-receptors and voltage-gated potassium channels (2). Propofol also potentiates glycine receptors which are are found all over the central nervous system and have a major role in regulating inhibition, e.g., in the brain stem (89).

Similar to extra-synaptic inhibition resulting from ambient GABA concentrations, the presence of ambient concentrations of glycine close to NMDA-receptors entails tonic depolarization. This tonic excitation diminishes the firing threshold of neurons and hence may counteract inhibition. The present work considers tonic inhibition only and neglects tonic excitation effect. Although it would be important to study tonic excitation effects, this additional study would exceed the major aim of the manuscript, namely demonstrating the fundamental effect of tonic anesthetic action.

In addition, by virtue of the focus on extra-synaptic action, the model proposed neglects known anesthetic effects on different receptors and ion channels, although they have been shown experimentally, e.g. (2, 90) and references therein, and theoretically (91) to affect EEG activity. Specifically, the latter work of Bojak et al. (91) considers anesthetic effects on hyperpolarization-activated cyclic nucleotidegated potassium channel 1 (HCN1) subunits which, effectively, increase the mean firing threshold in neural populations and strongly resembles the tonic inhibition induced by extra-synaptic GABA-receptors.

The model network topology includes a single module of a closed thalamo-cortical feedback loop (92) 364 365 comprising two thalamic nuclei and cortical excitatory and inhibitory neurons. This model represents a first approximation of brain networks since it neglects brain stem activity including the reticular activating 366 system (RAS) (93) which has significant modulating effects on attention, arousal and consciousness. 367 368 Future work will include structures of the brain stem, propofol action on glycine receptors, and will take into account the RAS - since its neural structures involved exhibit strong extra-synaptic inhibition (94, 95, 369 96). The model also neglects the cholinergic pathway originating from the basal forebrain (97) which is 370 known to co-regulate the level of consciousness (86). 371

Essentially, our theoretical model assumes population coding implying rate-coding response activity of neuron populations subjected to external thalamic noise. The model does not consider specific single neuron dynamics found experimentally under anesthetic conditions. For instance, it has been hypothesized that, at certain levels of anesthetic concentration, thalamic neurons switch their activity from tonic firing to bursting and induce loss of consciousness (98).

377

In spite of these limitations, our model reproduces qualitatively the action of propofol on EEG and reveals the possible impact of extra-synaptic GABAergic receptors on the EEG power. To our knowledge, the present work is the first to link extra-synaptic GABAergic action and experimental EEG. Future work will refine the model involving additional receptor action, e.g., tonic excitation caused by ambient glycine concentrations, and sub-cortical brain structures.

APPENDIX A. POWER SPECTRUM

383 The resting state (stationary state) of Eqs. (1) defined by $dV_a(t)/dt = 0$ obeys

$$\begin{aligned}
V_{E}^{*} &= \nu_{ee}S(V_{E}^{*}) + H_{b}\nu_{ei}S(V_{I}^{*}) + \nu_{es}S(V_{S}^{*}), \\
V_{I}^{*} &= \nu_{ie}S(V_{E}^{*}) + H_{b}\nu_{ii}S(V_{I}^{*}) + \nu_{is}S(V_{S}^{*}), \\
V_{S}^{*} &= \nu_{se}S(V_{E}^{*}) + H_{b}\nu_{sr}S(V_{R}^{*}) + \langle \phi_{N} \rangle, \\
V_{R}^{*} &= \nu_{re}S(V_{E}^{*}) + \nu_{rs}S(V_{S}^{*}),
\end{aligned}$$
(10)

384 where V_a^* denotes the resting state value of V_a for a = E, I, S, R. Moreover Eq. (3) gives $\phi_E^* = S(V_E^*)$. 385 Then we linearize Eqs. (1) about the obtained resting state and write them in a general matrix form of a 386 linear DDE as

$$\hat{\boldsymbol{L}}\left(\frac{\partial}{\partial t}\right)\boldsymbol{X}(t) = \boldsymbol{A}\boldsymbol{X}(t) + \boldsymbol{B}\boldsymbol{X}(t-\tau) + \boldsymbol{\xi}(t),$$
(11)

where

$$\boldsymbol{X}(t) = \begin{pmatrix} \phi_{E}(t) - \phi_{E}^{*} \\ V_{I}(t) - V_{I}^{*} \\ V_{S}(t) - V_{S}^{*} \\ V_{R}(t) - V_{R}^{*} \end{pmatrix}, \quad \hat{\boldsymbol{L}} = \begin{pmatrix} \tilde{\boldsymbol{L}} \frac{\tilde{\boldsymbol{D}}}{K_{11}} & \boldsymbol{0} & \boldsymbol{0} & \boldsymbol{0} \\ \boldsymbol{0} & \tilde{\boldsymbol{L}} & \boldsymbol{0} & \boldsymbol{0} \\ \boldsymbol{0} & \boldsymbol{0} & \tilde{\boldsymbol{L}} & \boldsymbol{0} \\ \boldsymbol{0} & \boldsymbol{0} & \boldsymbol{0} & \tilde{\boldsymbol{L}} \end{pmatrix},$$
$$\boldsymbol{A} = \begin{pmatrix} K_{1} & K_{2} & \boldsymbol{0} & \boldsymbol{0} \\ K_{4} & K_{5} & \boldsymbol{0} & \boldsymbol{0} \\ \boldsymbol{0} & \boldsymbol{0} & K_{8} \\ \boldsymbol{0} & \boldsymbol{0} & K_{10} & \boldsymbol{0} \end{pmatrix}, \quad \boldsymbol{B} = \begin{pmatrix} \boldsymbol{0} & \boldsymbol{0} & K_{3} & \boldsymbol{0} \\ \boldsymbol{0} & \boldsymbol{0} & K_{6} & \boldsymbol{0} \\ K_{7} & \boldsymbol{0} & \boldsymbol{0} & \boldsymbol{0} \\ K_{9} & \boldsymbol{0} & \boldsymbol{0} & \boldsymbol{0} \end{pmatrix}, \quad \boldsymbol{\xi}(t) = \begin{pmatrix} \boldsymbol{0} \\ \boldsymbol{0} \\ \sqrt{2\kappa}\boldsymbol{\xi}(t) \\ \boldsymbol{0} \end{pmatrix}, \quad (12)$$

387 with $\tilde{L} = (1 + i\omega/\alpha)(1 + i\omega/\beta_0)$, $\tilde{D} = (1 + i\omega/\gamma)^2$, and

$$K_{1} = \nu_{ee} , K_{2} = H_{b}\nu_{ei}\frac{\partial S_{I}[V]}{\partial V} |_{V=V_{I}^{*}} , K_{3} = \nu_{es}\frac{\partial S_{S}[V]}{\partial V} |_{V=V_{S}^{*}},$$

$$K_{4} = \nu_{ie} , K_{5} = H_{b}\nu_{ii}\frac{\partial S_{I}[V]}{\partial V} |_{V=V_{I}^{*}} , K_{6} = \nu_{is}\frac{\partial S_{S}[V]}{\partial V} |_{V=V_{S}^{*}},$$

$$K_{7} = \nu_{se} , K_{8} = H_{b}\nu_{sr}\frac{\partial S_{R}[V]}{\partial V} |_{V=V_{R}^{*}} , K_{9} = \nu_{re},$$

$$K_{10} = \nu_{rs}\frac{\partial S_{S}[V]}{\partial V} |_{V=V_{S}^{*}} , K_{11} = \frac{\partial S[V]}{\partial V} |_{V=V_{E}^{*}}.$$

388 The power spectral density matrix $P(\omega)$ of X(t) is the Fourier transform of the auto-correlation function 389 matrix $\langle X(t)^t X(t-T) \rangle$ (Wiener-Khinchine Theorem) leading to

$$\boldsymbol{P}(\omega) = 2\kappa\sqrt{2\pi}\tilde{\boldsymbol{G}}(\omega)\tilde{\boldsymbol{G}}^{t}(-\omega),$$

390 where $\tilde{G}(\omega)$ is the Fourier transform of the matrix Greens function and the high index t denotes the 391 transposed matrix (25).

At the end, the model assumes that excitatory activity generates the EEG and by virtue of the specific choice of external input to relay neurons, the power spectrum of the EEG depends just on one matrix component of the Greens function by

$$P_E(\omega) = 2\kappa\sqrt{2\pi}\tilde{G}_{1,3}(\omega)\tilde{G}_{1,3}(-\omega) = 2\kappa\sqrt{2\pi}\left|\tilde{G}_{1,3}(\omega)\right|^2,\tag{13}$$

395 cf. Eq. (8), where

$$\tilde{\boldsymbol{G}}(\omega) = \frac{1}{\sqrt{2\pi}} [\hat{\boldsymbol{L}} - \boldsymbol{A} - \boldsymbol{B}e^{-i\omega\tau}]^{-1} = \frac{1}{\sqrt{2\pi}} \begin{bmatrix} \tilde{L}\frac{\tilde{D}}{K_{11}} - K_1 & -K_2 & -K_3e^{-i\omega\tau} & 0\\ -K_4 & \tilde{L} - K_5 & -K_6 & 0\\ -K_7e^{-i\omega\tau} & 0 & \tilde{L} & -K_8\\ -K_9e^{-i\omega\tau} & 0 & -K_{10} & \tilde{L} \end{bmatrix}^{-1}.$$
 (14)

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APPENDIX B. PARAMETER VALUES

The subsequent table gives the nominal values of model parameters.

Parameter	Symbol	Nominal value
Maximum firing-rate of all populations	S^{max}	250 Hz
Mean firing threshold of all populations	θ	15 mV
Firing rate variance of all populations	σ	10 mV
Type-I population effect constant of all populations	ρ	$0.08 \ { m mV}^{-1}$
Synaptic rise rate	α	$200 \ s^{-1}$
Synaptic decay rate	β_0	$50 \ s^{-1}$
Synaptic strength from E to E neurons	ν_{ee}	1.2 mVs
Synaptic strength from E to I neurons	$ u_{ie}$	1.2 mVs
Synaptic strength from E to S neurons	ν_{se}	1.2 mVs
Synaptic strength from E to R neurons	$ u_{re}$	0.4 mVs
Synaptic strength from I to I neurons	$ u_{ii}$	-1.8 mVs
Synaptic strength from I to E neurons	$ u_{ei}$	-1.8 mVs
Synaptic strength from S to E neurons	ν_{es}	1.2 mVs
Synaptic strength from S to I neurons	$ u_{is}$	1.2 mVs
Synaptic strength from S to R neurons	v_{rs}	0.2 mVs
Synaptic strength from R to S neurons	ν_{sr}	-0.8 mVs
Mean value of external input	$\langle \phi_N \rangle$	1 mV
Intensity of external noise	κ	0.1 mV
Cortical damping rate	γ	$150 \ s^{-1}$
Transmission delay between cortex and thalamus	au	40 ms

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DISCLOSURE/CONFLICT-OF-INTEREST STATEMENT

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

AUTHOR CONTRIBUTIONS

399 MH has chosen the model and performed the model analysis, AH has conceived the study and JS has acquired and analyzed the experimental data. The authors have written the manuscript to equal parts and have approved the final version.

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