Title: Lasting modulation of in-vitro oscillatory activity with weak direct current stimulation

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Lasting modulation of \textit{in-vitro} oscillatory activity with weak direct current stimulation

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\textbf{Abstract}

Transcranial Direct Current Stimulation (tDCS) is emerging as a versatile tool to affect brain function. While acute neurophysiological effects of stimulation are well understood, little is know about the long term effects. One hypothesis is that stimulation modulates ongoing neural activity which then translates into lasting effects via physiological plasticity. Here we used carbachol-induced gamma oscillations in hippocampal rat slices to establish whether prolonged constant current stimulation has a lasting effect on endogenous neural activity. During 10 minutes of stimulation, power and frequency of gamma oscillations, as well as multi-unit activity were modulated in a polarity specific manner. Remarkably, the effects on power and multi-unit activity persisted for more than 10 minutes after stimulation terminated. Using a computational model we propose that altered synaptic efficacy in excitatory and inhibitory pathways could be the source of these lasting effects. Future experimental studies using this novel \textit{in-vitro} preparation may be able to confirm or refute the proposed hypothesis.

\textbf{Introduction}

The number of studies on transcranial direct current stimulation (tDCS) has rapidly increased in recent years (Brunoni et al., 2012). Human studies have shown improvements in behavioral and cognitive performances after transcranial stimulation (Foerster et al., 2012; Javadi et al., 2011). Pharmacological interventions in human studies point to possible synaptic changes as well as changes in neuromodulator release as the cause for the after-effects of the stimulation (Nitsche and Paulus, 2000; Nitsche et al., 2003) (for a review, Stagg and Nitsche, 2011). A few animal studies have also shown effects that outlast the stimulation period using evoked response (Márquez-Ruiz et al., 2012; Cambiaghi et al., 2010;...
Fritsch et al., 2010; Ranieri et al., 2012; Gartside, 1968). However, how these results may relate to the effects measured in human studies is not fully understood.

In recent years, brain oscillations – rhythmic neuronal activity reflecting coherent spiking – have become a target for transcranial electrical stimulation (Herrmann et al., 2013). Gamma oscillations in the range of 25-100 Hz, for example, are ubiquitous in the brain (Buzsáki and Draguhn, 2004), play a role in neuronal coding (Fries et al., 2007; Wang, 2010) and have been associated with attention and memory in humans (Jensen et al., 2007). Gamma rhythms in the 30 Hz range synchronize hippocampal activity during memory replay (Carr et al., 2012) and reduced gamma power leads to impaired spatial working memory and exploratory behavior (Fuchs et al., 2007). Therefore, modulation of gamma rhythms with weak currents could potentially affect brain function. While many human studies used transcranial alternating current stimulation (tACS), only few considered how DC stimulation could affect brain oscillations (Antal et al., 2004; Polanía et al., 2011).

The acute effects of DC stimulation on single neurons and networks of neurons have been extensively characterized in animals (Bikson et al., 2004; Fröhlich and McCormick, 2010; Reato et al., 2010). However, no animal studies have shown lasting effects of DC stimulation on brain oscillations. Here, we present an in-vitro model of gamma oscillations and DC stimulation that shows lasting effects of stimulation.

Oscillations in the low-gamma frequency range can be reliably induced in hippocampal slices using carbachol, a cholinergic agonist (Fisahn et al., 1998). We have shown previously that weak electric fields can modulate the magnitude of these in-vitro oscillations acutely in an amplitude and frequency specific manner (Reato et al., 2010). Here we use the same slice model and report that modulation of gamma oscillations with weak constant electric fields applied for a prolonged time (10 minutes) outlasts the period of stimulation. Based on simulations with a previously validated computational model we propose that the after-effects of stimulation are the result of balanced changes in excitatory and inhibitory synaptic strength.

Materials and Methods

Hippocampal slice recordings: Slice preparation and recordings from male 3-5 weeks old Wistar rats (CCNY-IACUC, protocol 0846) followed the procedures of (Reato et al., 2010). Extracellular recordings were performed using 3 electrodes per slice placed in the CA3c stratum pyramidale area of the hippocampus at the distance of $250\ \mu m$. Perfusion with carbachol (20 $\mu M$) started 5 minutes after placing the glass pipettes and the recordings lasted for the following 2 hours without moving the electrodes. Recordings were aligned in time across slices using the beginning of carbachol perfusion.

Electrical field stimulation: Spatially-uniform electric fields were applied to slices with varying amplitudes by passing current between two parallel Ag-AgCl wires placed in the ACSF across the slice (Gluckman et al., 1996; Bikson et al., 2004). All experimental results are reported as a function of this electric field magnitude. Slices were aligned
in the chamber such that the induced uniform electric field was parallel to CA3c pyramidal neurons (Fig. 1A). Before every recording, electric fields were calibrated by passing current through the field wires and measuring the corresponding voltages between them (representative voltage measurements in Fig. 1A). Note that a linear voltage indicates a uniform value of the electric field. “Positive” (anodal) field polarity was defined as the positive electrode on the CA1 alveus side, and “negative” (cathodal) field polarity was defined as negative on the CA1 alveus side (Bikson et al., 2004). Weak positive field stimulation is thus typically expected to depolarize CA3 pyramidal cell somata, while weak negative field stimulation should hyperpolarize CA3 somata (Deans et al., 2007). DC stimulation commenced 60 minutes after the recordings started (55 minutes after application of carbachol). During this time, the oscillations largely stabilized in power and frequency, although a continued drift in power is often evident (Fig. 1C). Stimulation was applied for 10 minutes with amplitudes of -20 V/m (n = 6), -10 V/m (n = 24), 0 V/m (control, n = 26), +10 V/m (n = 23) and +20 V/m (n = 8). Each slice was stimulated with a single stimulation intensity.

Power and frequency analysis: Gamma power and frequency were estimated using multitaper spectral analysis. Power was computed in 1 minute segments using the Chronux toolbox (http://chronux.org/, Mitra and Bokil, 2008) with a time-bandwidth product of $WT = 2$ and using 3 tapers. The frequency range of carbachol-induced activity was then detected semi-automatically for each slice (location of the peak $\pm 2\sigma$ but manually reduced to exclude, if present, electric noise contaminating that frequency range) and mean-power was calculated for that frequency range. Gamma power was averaged across electrodes provided it was at least 5 dB above noise as compared to the first 5 minutes of recording. To compare across slices, these mean values were normalized by the pre-stimulation gamma power (50-60 min). Gamma peak-frequency was estimated from the multitaper analysis for each 1 minute segment by considering the location of the peak-power in the gamma band.

Multi-unit activity detection: Multi-unit activity (MUA) was detected by thresholding the extracellular recordings after high-pass filtering (300 Hz cut-off frequency). The value of the threshold for automatic units detection was set to $7 \cdot \text{median} \left( \frac{|x|}{0.6745} \right)$ (Quiroga et al., 2004) where $x$ was the high-pass filtered extracellular signal during the first 5 minutes of recording (before carbachol perfusion). A 1 ms dead time for detection was used. Our main results do not depend strongly on the specific threshold for MUA detection. A high threshold for detection, as we used here, was chosen to decrease false positive detections possibly due to electric artifacts (note that in this study we used conventional electrodes for LFP recordings). This simple method allowed to easily estimate MUA changes in our different stimulation conditions. Coherence between the candidate units and the extracellular local field potentials (LFP) was estimated using Chronux (Mitra and Bokil, 2008) (http://chronux.org/). Only electrodes that showed a strong unit-to-field-potential coherence ($> 0.3$) in our frequencies of interest were considered for further analysis. When multiple electrodes detected MUA, the frequency of events was averaged across these electrodes in 1 minute temporal windows. To compare across slices, estimated average rates (events per second) before the stimulation (50-60 min) where subtracted from each trace. Note
that since the recordings started before the emergence of coherent activity in the slice and the electrodes were not moved throughout the experiment, a good MUA signal (high coherence with the LFP) was not always detected. Therefore, the number of slices for LFP and MUA analysis differ in the Results section.

**Sensitivity of power and multi-unit activity to electrical stimulation:** We measured the sensitivity of the network oscillatory power to the applied field. In equations, $\Delta P = g_p E$, where $g_p$ represents how much power changes (in dB) per V/m electric field applied. MUA modulation follows a similar equation, $\Delta R = g_r E$, where $g_r$ indicates how many Hz the estimated rate changes with stimulation intensity. These sensitivities were estimated with a linear fit as a function of the stimulation intensities using all available slices. Non-parametric statistics were obtained by randomizing the stimulation amplitudes and performing the linear fits to this random data. p-values were then computed using these shuffled statistics. Statistical tests were performed for the 10 minutes of stimulation and the subsequent 10 minutes. This was based on previous literature showing that excitability changes in motor cortex outlast the stimulation period for durations comparable to the stimulation period (5 minutes in (Nitsche and Paulus, 2000), and 20 minutes in (Nitsche and Paulus, 2001).

**Computational model:** Modeling of gamma oscillations induced by carbachol and its response to electric stimulation follows the methods of (Reato et al., 2010). Briefly, the voltage behavior of single neurons is captured by Izhikevich’s single-compartment neuron model (Izhikevich, 2003). Here the network consists of 800 excitatory and 200 inhibitory neurons synaptically connected with all-to-all connections and 40% sparseness. Gamma oscillations in the model are generated by the interplay of increased excitation (simulating the effect of carbachol, Fisahn et al., 2002) and fast inhibitory feedback (Bartos et al., 2007). Weak electrical stimulation was implemented as a low-pass filtered current that polarizes all pyramidal neurons, according to experimental data (Deans et al., 2007). The parameters of the models are set such that 1 V/m electric field induces a polarization of about 0.1 mV, consistent with previous studies (Radman et al., 2007; Deans et al., 2007; Bikson et al., 2004; Fröhlich and McCormick, 2010).

Here we tested how changes in the strength of synaptic connections affect power, frequency and firing rate of excitatory neurons during simulated gamma oscillations. The form of synaptic connections in our model is $w_{xx} = \tilde{w}_{xx} + [0, k_{xx}\tilde{w}_{xx}]$ if the connection is excitatory or $w_{xx} = \tilde{w}_{xx} + [k_{xx}\tilde{w}_{xx}, 0]$ if inhibitory, and where $xx = \{ee,ei,ie,ii\}$ indicates the type of connection (excitatory to excitatory, excitatory to inhibitory, inhibitory to excitatory, inhibitory to inhibitory). $\tilde{w}_{xx}$ represents the baseline value of the connection, $\tilde{w}_{xx}$ the maximum value of the uniform distribution (from 0 to $\tilde{w}_{xx}$) and $k_{xx}$ is a parameter that has been changed here to simulate changes in synaptic connections. As in Reato et al., 2010, we used here $\tilde{w}_{ee} = \tilde{w}_{ei} = 0$, $\tilde{w}_{ie} = -0.8$, $\tilde{w}_{ii} = -0.3$ and $\tilde{w}_{ee} = 0.65$, $\tilde{w}_{ei} = 2$, $\tilde{w}_{ie} = -0.9$, $\tilde{w}_{ii} = -0.8$. Throughout the text, the expression “balanced excitation/inhibition” indicate an equal level of excitatory and inhibitory inputs on single neurons during the gamma cycle. This definition is based on previous literature indicating per-neuron balanced excitatory and inhibitory current during gamma oscillations (Atallah and Scanziani, 2009) and slow-waves (Haider et al., 2006; Shu
Results

Extracellular recordings were performed with multiple electrodes located in the CA3c region of rat hippocampal slices (n = 87, Fig. 1A). Carbachol was perfused continuously beginning 5 minutes after the start of recording. Carbachol induced strong gamma oscillations (25-35 Hz, 5-25 dB over noise) that emerge ~20 minutes after starting the perfusion, consistent with other studies (Colgin et al., 2003). In the average across slices, the oscillations became relatively stable in power and frequency after ~60 minutes (average spectrogram in Fig. 1B, n = 26). Gamma power and multi-unit activity (MUA) were measured over the whole duration of the recordings (2 hours) and were normalized by their average value before the stimulation. Gamma power and MUA were not statistically different across the five stimulation conditions before the stimulation (ANOVA, n = 87, p = 0.61 for power and n = 39, p = 0.77 for MUA). 55 minutes after the start of carbachol perfusion, slices were electrically stimulated with constant electric fields for 10 minutes. For each slice, only one stimulation intensity was used: -20 V/m (n = 6), -10 V/m (n = 24), 0 V/m (n = 26), +10 V/m (n = 23), +20 V/m (n = 8). Figs. 1C-D show average traces of gamma power and MUA in the different stimulation conditions (the stimulation starts at 0 minutes). The significant variability observed over the recording period for individual traces required recording of a large number of slices (n=87).

We tested whether electric fields modulated gamma oscillations and MUA in an intensity-dependent manner. We grouped the data from all the slices (n = 87 for power, and n = 39 for MUA) and performed a linear regression as a function of stimulation intensity for the 10 minute interval during the stimulation (acute effects) and after stimulation (persisting effects). Combining data from all stimulation conditions was necessary in order to average out the strong fluctuation observed over time and across slices. The first minute in both intervals was excluded from the analysis to avoid possible transients or artifacts resulting from turning the stimulator on/off. Gamma power was significantly modulated during the stimulation (n = 87, estimated slope of the linear fit, \( g_p = (0.03 \pm 0.01) \text{ dB/(V/m)} \), \( p = 0.001 \), Fig. 2A left) and a similar effect was also measured for MUA (n=39, estimated slope of the linear fit, \( g_r = (0.11 \pm 0.03) \text{ Hz/(V/m)} \), \( p=3 \times 10^{-6} \), Fig. 2B left). The positive offset of the regression lines reflect the continuous strengthening of gamma oscillations even 55 minute after carbachol perfusion. A positive slope implies that both gamma power and MUA are higher when positive (anodal) electric fields are applied and lower for negative (cathodal) fields. Importantly, the effects outlasted the stimulation in the subsequent 10 minutes for both power (n=87, estimated slope of the linear fit, \( g_p = (0.02 \pm 0.01) \text{ dB/(V/m)} \), \( p = 0.02 \), Fig. 2A right) and MUA (n = 39, estimated slope of the linear fit, \( g_r = (0.10 \pm 0.03) \text{ Hz/(V/m)} \), \( p=0.001 \), Fig. 2B right). The same analysis performed on frequency changes did not reveal any significant effects of fields (\( p=0.1 \) during and 0.3 after stimulation). To determine the exact progression of power and MUA changes, we then estimated power and MUA sensitivity to the electric field (considering data from all the slices, as previously done for Fig. 2A-B)
resolved in 1 minute segments (Fig. 2C-D). Both power and MUA sensitivity continuously increased during stimulation (dark gray shading) and then decayed after the end of the stimulation (light gray shading). The increasing confidence intervals reflect the substantial variability of the gamma oscillations (light gray area, 5% and 95% estimated by shuffle statistics). The time courses of power and MUA are strongly correlated ($r^2 = 0.65$). Taken together, these results show that weak electrical stimulation can affect gamma oscillations and MUA and that the effects outlast the stimulation for at least 10 min.

Next, we tried to determine possible causes for these lasting effects using an experimentally-validated computational model for carbachol-induced gamma oscillations and their response to electric field stimulation (Reato et al., 2010). In the model, increasing field magnitudes leads to a monotonic increase of gamma power and firing rate (Fig. 3A-B). Interestingly, increasing field intensity also leads to a monotonic decrease in gamma frequency (Fig. 3C) that correlates with changes in gamma-power ($r^2 = 0.85$, Fig. 3D). This relationship was confirmed with our present in-vitro data during the stimulation (Fig. 3E, n=87, $p=6 \times 10^{-4}$) and is consistent with previous data showing that a balance between excitation and inhibition is the cause of this linear relationship (Atallah and Scanziani, 2009). In the computational model the relationship between electric field magnitude and gamma frequency is non-linear (Fig. 3C). Thus, we analyzed the changes in gamma peak-frequency in the in-vitro data separately for the different stimulation conditions and find significant effects in particular for negative field stimulation (Fig. 3F). To summarize, our experimental results showed acute effects for gamma power, frequency and MUA, pointing to a balanced modulation of excitatory and inhibitory activity, yet persistent effects were only evident for oscillatory power and MUA.

We then used the computational model to investigate whether the observed lasting changes could be explained by changes in synaptic strength, notably a change in gamma power and MUA but not frequency. We focused on synaptic changes, as they are often assumed to underlie the persistent effects observed in human studies (see Discussion). We modulated the strength of excitatory-to-excitatory ($e\rightarrow e$) synaptic connections, as well as the strength of the inhibitory feedback (excitatory-to-inhibitory, $e\rightarrow i$, or inhibitory-to-excitatory, $i\rightarrow e$, synaptic strength). Modulation of inhibitory synaptic strength was motivated by recent evidence for plasticity in inhibitory pathways (Kullmann et al., 2012). Modulating the strength of $e\rightarrow e$ synapses in the model did not strongly modulate the power of the oscillations (Fig. 4A-D), while changing $e\rightarrow i$ or $i\rightarrow e$ synapses strongly modulated gamma power (Fig. 4A,D). Both phenomena have previously been observed experimentally for endogenous neocortical gamma activity (Morita et al., 2008; Sohal et al., 2009). This provides further confidence in the present computational model. Here gamma frequency depends on both $e\rightarrow e$ and $e\rightarrow i$ connections but less on $i\rightarrow e$ (Fig. 4B-E). Firing rate is more sensitive to changes in $e\rightarrow e$ connections (Fig. 4C-F). Points in this parameter space that are consistent with the present experimental observation are indicated with an “0” for sham stimulation and “+”, “-” for positive and negative field stimulation. Along these diagonals, power and firing-rate change, but not oscillation frequency. Therefore, the computational results suggest that the observed lasting changes may be
explained by a lasting modulation of excitation matched by a corresponding change in inhibitory feedback.

**Discussion**

Transcranial electrical stimulation is a versatile tool to modulate brain activity (Nitsche and Paulus, 2000; Fregni et al., 2006; Fecteau et al., 2007; Fridriksson et al., 2011). *In-vivo* and *in-vitro* studies have demonstrated that electric fields, whose amplitude is comparable to the one expected in tDCS, can modulate firing rate (Chan and Nicholson, 1986), spike timing (Radman et al., 2007) and the magnitude of synaptic responses (Kabakov et al., 2012; Rahman et al., 2013). We have previously shown that acute effects of weak electrical stimulation can be amplified during endogenous oscillatory activity (Reato et al., 2010, 2013). These results suggest that brain oscillations may be a sensitive target for transcranial electrical stimulation with constant currents. Previous *in-vitro* and *in-vivo* studies have only shown acute effects of stimulation on oscillatory activity (Reato et al., 2010; Fröhlich and McCormick, 2010; Ali et al., 2013; Ozen et al., 2010) and there are few reports on longer term effects in human studies (see Antal et al., 2004; Polanía et al., 2011). The majority of studies on oscillatory activity have used alternating current stimulation with the goal of enhancing brain oscillations (Marshall et al., 2006; Pogosyan et al., 2009; Kirov et al., 2009; Zaehle et al., 2010; Santarnecchi et al., 2013; Helfrich et al., 2014).

Here we found that weak constant current electrical stimulation applied for a longer period of time can induce lasting effects, measurable as altered gamma power and multi-unit activity. These lasting effects cannot be explained as persistent network activity in the absence of some adaptive process since in our previous work gamma power returned to baseline activity within 100 ms after short-lasting DC field stimulation (Reato et al., 2010). Importantly, the after-stimulation effect was consistent with the acute effect, reminiscent of Hebbian or activity-dependent plasticity and contrary to homeostatic plasticity (Fricke et al., 2011; Reato et al., 2013).

The field intensities used in this study are above those predicted to occur during tDCS, estimated to be maximum 1 V/m using conventional electrode montages (Datta et al., 2009; Ozen et al., 2010). These currents only induce a small polarization of the membrane (maximum 0.2 mV per V/m), which cannot lead to action potentials in quiescent neurons (Bikson et al., 2004). Previous *in-vitro* studies have shown that for such low intensity fields (subthreshold), most of the acute effects scale linearly with the change in field amplitude (Bikson et al., 2004; Deans et al., 2007; Reato et al., 2010), including changes in synaptic response (~1% per V/m applied; Rahman et al., 2013). Therefore, the sensitivities observed here may also scale linearly with the field intensities. In this context we note that several factors may make the human brain more susceptible to electric fields, including larger sensitivity of individual neurons (due to size; Radman et al., 2009) and higher number of synaptic connections compared to our *in-vitro* preparation (sensitivity to fields may increase with the number of synaptic inputs a neuron receives; Reato et al., 2013). Either way, our field amplitudes are still much below those generated with transcranial magnetic stimulation (TMS; Pascual-Leone et al., 2002) or deep brain stimulation.
(DBS; Perlmutter and Mink, 2006), estimated in the order of 100 V/m (Salinas et al., 2009).

To generate a hypothesis for the possible cause of the experimental results we turned to computational modeling. The model we used matches key features of weak-field stimulation on carbachol-induced gamma oscillations (Reato et al., 2010). Specifically, the model matches the firing properties of excitatory and inhibitory neurons and their timing within the gamma cycle (Hájos et al., 2004; Oren et al., 2006). The model successfully predicts firing rate and spike timing changes during AC or DC field stimulation *in-vitro*. Without further modifications, this model reproduced the correlation observed in the present experiment between power and frequency changes due to DC stimulation. A variant of this model also successfully predicted the effects of weak transcranial stimulation on slow-waves oscillations *in-vivo* (Ali et al., 2013). More complex models that capture physiological details such as gap-junctions or the role of different neuronal compartments (Tiesinga et al., 2001; Traub et al., 2000) were not necessary to replicated the relevant experimental findings. Using the model, we focused on modulation of synaptic connections because tDCS is thought to modulate concentrations of neurotransmitters and neuromodulators (Stagg and Nitsche, 2011; Nitsche et al., 2012), which in turn are known to affect synaptic efficacy. Based on the computational model we hypothesize that in gamma-networks, weakly depolarizing electric fields lead to a balanced increase of excitatory and inhibitory synaptic currents.

The lasting effects we measured experimentally could be mediated by a number of cellular mechanisms, which we discuss below.

**Brain-derived neurotrophic factor (BDNF):** It has been shown in humans (Antal et al., 2010) and *in-vitro* (Fritsch et al., 2010) have shown that the lasting effects of weak electrical stimulation can be mediated by BDNF release. BDNF release is activity-dependent (Park and Poo, 2013) with a self-reinforcing feedback-loop involving acetylcholine (Knipper et al., 1994). Thus, an acute increase in gamma activity due to electrical stimulation could be further enhanced by increased BDNF release, and this enhancement should outlast stimulation because of the longer time scale of BDNF release (Aicardi et al., 2004). Interestingly, BDNF affects both excitatory and inhibitory neurons (Park and Poo, 2013) via the TrkB receptor, which has also been implicated in gamma activity (Zheng et al., 2011). Thus, any enhancing effect resulting from increased BDNF release may strengthen both excitation and inhibitory feedback, as we have hypothesized here.

**Acetylcholine:** Carbachol activates acetylcholine receptors leading to increased neuronal activity in hippocampus. It is well established that acetylcholine can induce hippocampal plasticity (Drever et al., 2011; Galey et al., 1994; Markevich et al., 1997; Fernández de Sevilla et al., 2008). Indeed, carbachol alone can induce lasting effects on the acetylcholine receptors (Auerbach and Segal, 1994) and can facilitate hippocampal LTP (Auerbach and Segal, 1996). Moreover, carbachol increases network responsiveness to external stimuli *in-vivo* (Rodriguez et al., 2004; Rasmusson, 2000) and can induce lasting effects on evoked responses (Rodriguez et al., 2004; Bröcher et al., 1992), presumably by increasing precision of spike timing in the network. Finally, long lasting effects on cortical activity can be induced when sensory stimulation is paired with activation.
of cholinergic inputs from the basal forebrain in-vivo (Froemke et al., 2013). It is thus possible that the increased activity due to electric fields in the presence of carbachol is translated also into increased carbachol-induced plasticity. Interestingly, a recent in-vivo study reported that acetylcholine mediated-learning induces strengthening at both excitatory and inhibitory synapses (Mitsushima et al., 2013), supporting our hypothesis that weak electrical stimulation may affect both types of synapses.

**Spike-timing dependent plasticity (STDP):** The altered gamma activity with firing periods in the order of 10-30 ms may induce NMDA-mediated STDP (Wespatat et al., 2004). Thus, increased firing due to field stimulation could lead to altered synaptic efficacies via STDP which outlast the period of stimulation.

**Membrane excitability:** Finally, stimulation may affect membrane excitability (non-synaptic, Ardolino et al., 2005). For example, stimulation-induced slow changes in neuromodulator release could lead to slow changes in neuronal membrane properties giving rise to changes in the population dynamics and the studied after-effects (Augustin et al., 2013). Considering the nature of gamma oscillations in hippocampus, increase/decrease in excitability of excitatory neurons (the most affected by electrical stimulation, Radman et al., 2009) could also lead to a balanced increase/decrease of inhibitory feedback.

In all instances, we propose that stimulation acutely affects ongoing activity, which then leads to lasting effects via endogenous plasticity mechanisms. We argue that using our slice preparation we will be able to test our specific hypothesis that balanced synaptic changes mediate the effects of weak electric fields on gamma oscillations.

**References**


Figure 1: Carbachol-induced gamma oscillations during electrical stimulation. A: Extracellular recordings were performed in the CA3c area of rat hippocampal slices. Spatially uniform DC electric fields were applied using AgCl wire electrodes in the bath. Pseudo-colors represent the voltage as recorded before a typical session. Linear voltage means constant electric field across the slice. B: Average spectrogram of carbachol-induced gamma oscillations for control condition (n = 26 slices). C: Average traces of gamma power in five different stimulation conditions (-20 V/m, -10 V/m, 0 V/m, +10 V/m, +20 V/m, mean ± SEM). Shaded gray rectangle indicates the stimulation period (10 minutes). Average traces for control condition (green line) are reported in each figure for direct comparison. D: Average traces of multi-unit activity during gamma oscillations in the same five stimulation conditions (mean ± SEM). Shaded gray rectangle indicates the stimulation period (10 minutes). Average traces for control condition (green line) are reported in each figure for direct comparison.

Figure 2: Modulation of gamma power and MUA by weak electrical stimulation. A: Gamma power changes for each slice during (left, immediate) and after (right, persisting) the application of electrical stimulation (n=87). Black line represents a linear fit. B: MUA changes for each slice during (left, acute) and after (right, plastic) the application of electrical stimulation (n=39). C: Sensitivity of gamma power to the applied field as a function of time (black curve). D: Sensitivity of MUA to the applied field as a function of time (black curve). The dark gray rectangles indicate the stimulation period, the light gray rectangles indicate the interval considered for persisting effects. Gray shading represents 5% and 95% confidence interval.

Figure 3: Effects of electrical stimulation on gamma power, peak frequency (frequency of the peak power) and firing rate. A: Gamma power change as a function of electric field amplitude in the computational model. B: Average firing rate change of excitatory neurons a function of electric field amplitude in the computational model. C: Peak gamma frequency change as a function of electric field amplitude in the computational model. Results indicate the average of 20 simulations (gray shading represents standard deviation). D: Acute gamma power changes as a function of frequency changes in the model. Points reflect repeated simulations and stimulation amplitudes (as in A-C). E: Acute gamma power changes as a function of frequency changes during the stimulation in the experimental data. Each point represents a slice (n=87) in the different conditions. F: Frequency changes during the stimulation in the experimental data for the different conditions. ** p=0.01.
Figure 4: Gamma power, frequency and average firing rate changes as a function of changes of the strength of synaptic connections in the computational model. **A-B-C:** Gamma power, frequency and firing rate changes as a function of changes in excitatory-to-excitatory ($k_{ee}$) and excitatory-to-inhibitory ($k_{ei}$) synaptic connections. **D-E-F:** Gamma power, frequency and firing rate changes as a function of changes in excitatory-to-excitatory ($k_{ee}$) and inhibitory-to-excitatory ($k_{ie}$) synaptic connections. "0" indicates points in the parameter space corresponding to sham condition, while "+" and "-" indicate positive (depolarizing) or negative (hyperpolarizing) stimulation.
A

B

C

D

Stimulation ON

0 V/m, n = 26

-20 V/m, n = 6

-10 V/m, n = 24

+10 V/m, n = 23

+20 V/m, n = 8

0 V/m, n = 9

-20 V/m, n = 4

-10 V/m, n = 12

+10 V/m, n = 11

+20 V/m, n = 3

Time (min)

Power change (dB)

MUA change (Hz)
A) **Power**

Immediate: $p = 0.001$

Persisting: $p = 0.02$

B) **MUA**

Immediate: $p = 3e-6$

Persisting: $p = 0.001$

C) **Power sensitivity**

Immediate: $p = 0.1$

Persisting: $p = 0.05$

D) **MUA sensitivity**

Immediate: $p = 0.1$

Persisting: $p = 0.05$
**Figure Legends**

A. Power change (dB) vs. Electric field (V/m)

B. Firing rate change (Hz) vs. Electric field (V/m)

C. Peak frequency change (Hz) vs. Electric field (V/m)

D. Power change (dB) vs. Peak frequency change (Hz)

E. Power change (dB) vs. Peak frequency change (Hz)

F. Peak frequency change (Hz) vs. Electric field (V/m)

*p = 6e-4*