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RESEARCH ARTICLE

Computational Neuroscience

Contrast detection is enhanced by deterministic, high-frequency transcranial alternating current stimulation with triangle and sine waveform

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Abstract

Stochastic resonance (SR) describes a phenomenon where an additive noise (stochastic carrier-wave) enhances the signal transmission in a nonlinear system. In the nervous system, nonlinear properties are present from the level of single ion channels all the way to perception and appear to support the emergence of SR. For example, SR has been repeatedly demonstrated for visual detection tasks, also by adding noise directly to cortical areas via transcranial random noise stimulation (tRNS). When dealing with nonlinear physical systems, it has been suggested that resonance can be induced not only by adding stochastic signals (i.e., noise) but also by adding a large class of signals that are not stochastic in nature that cause "deterministic amplitude resonance" (DAR). Here, we mathematically show that high-frequency, deterministic, periodic signals can yield resonance-like effects with linear transfer and infinite signal-to-noise ratio at the output. We tested this prediction empirically and investigated whether nonrandom, high-frequency, transcranial alternating current stimulation (tACS) applied to the visual cortex could induce resonance-like effects and enhance the performance of a visual detection task. We demonstrated in 28 participants that applying 80-Hz triangular-waves or sine-waves with tACS reduced the visual contrast detection threshold for optimal brain stimulation intensities. The influence of tACS on contrast sensitivity was equally effective to tRNS-induced modulation, demonstrating that both tACS and tRNS can reduce contrast detection thresholds. Our findings suggest that a resonance-like mechanism can also emerge when deterministic electrical waveforms are applied via tACS.

NEW & NOTEWORTHY Our findings extend our understanding of neuromodulation induced by noninvasive electrical stimulation. We provide the first evidence showing acute online benefits of transcranial alternating current stimulation (tACS)_{triangle} and tACS_{sine} targeting the primary visual cortex (V1) on visual contrast detection in accordance with the resonance-like phenomenon. The "deterministic" tACS and "stochastic" high-frequency-transcranial random noise stimulation (tRNS) are equally effective in enhancing visual contrast detection.

contrast sensitivity; deterministic amplitude resonance; high-frequency transcranial alternating current stimulation; neuromodulation; stochastic resonance

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INTRODUCTION

On Stochastic Resonance

Stochastic resonance (SR) was discovered in the context of the hysteresis features of climate (ice age) (1–3). Since then it has been generalized and studied in a variety of naturally occurring processes including biological systems (4, 5). Demonstrations of SR in the nervous system were carried out on crayfish mechanoreceptors (6), neurons in crickets (7), mice (8, 9), rats (10–12), cats (13), and humans (see Stochastic Resonance Effects on Neural Processing) with studies consistently reporting enhanced system performance. Signal enhancement is described in a vast body of literature as the basic property of resonance mechanisms (14). Here, we survey a few basic features of SR that are directly relevant to our paper. In general, the quality of signal transfer through a system is characterized by the following parameters at the output: amplification (or the signal strength), linearity, signalto-noise ratio, and phase shift.

SR is a phenomenon where the transfer of a periodic or aperiodic signal in a nonlinear system is optimized by an additive—typically Gaussian—noise (15). Note that originally, when SR was studied in binary systems, it represented a frequency-resonance, that is, matching the period time of the periodic signal with the mean residence time in the potential wells of the binary system driven by a stochastic carrier-wave (noise). Later the argument behind the name SR was modified to amplitude-resonance. Today, "resonance" means an optimal root-mean-square (RMS) amplitude value of the noise, i.e., amplitude-resonance at the carrier-wave RMS amplitude level for the best signal transmission.

In the initial phase of SR research, the nonlinear systems were bistable (1). At a later stage, it was discovered that monostable systems (including neurons) also offer SR (16). Moreover, it was realized that the memory/hysteresis effects of the bistable systems actually cause a stochastic phase shift (phase noise) that negatively impacts the quality of the transferred signal (17). Due to this fact, the best stochastic resonators are the memory-free threshold elements (TE), such as the level crossing detector (LCD) (18) and the comparator (19). The LCD device (the simplest model of a neuron) produces a short, uniform spike whenever its input voltage amplitude is crossing a given threshold level in a chosen, typically positive direction. On the other hand, the comparator has a steady binary output where the actual value is dictated by the situation of the input voltage amplitude compared with a given threshold level: for example, in the subthreshold case the output is "high" whereas in the suprathreshold case, it is "low."

At the output of a stochastic resonator, the signal strength (SS), the signal-to-noise-ratio (SNR), the information entropy, and the Shannon information channel capacity show maxima versus the intensity of the additive input noise. However, these maxima are typically located at different noise intensities. Exceptions are the SNR and information entropy that are interrelated by a monotonic function; thus they have the same location of their maxima, see the arguments relevant for neural spike trains (20). On the other hand, the information channel capacity of SR in an LCD and neural spike trains has the bandwidth

as an extra variable controlled by the input (the higher the input noise the higher the bandwidth); thus the different location of its maximum is at higher input noise than for the maximum of the SNR (21).

It is important to note that, in the linear response limit, that is, when the input signal is much smaller than the RMS amplitude of the additive carrier-wave (Gaussian noise), the SNR at the output is always less than at the input (see the mathematical proof in Ref. 15). Consequently, the information content at the output is always less than at the input. On the contrary, in the nonlinear response limit, the SNR at the output can be enhanced by several orders of magnitude compared with its input value provided the signal has a small duty cycle, such as neural spikes do (17, 22). Yet, due to the unavoidable noise at the output, which is the unavoidable impact of the stochastic carrier wave (noise), the information at the output (signal plus noise) is always less than in the original input signal without the added carrier wave (noise).

Therefore, if a proper additive, high-frequency, periodic time function could be used as carrier wave in a stochastic resonator instead of Gaussian noise, the fidelity and the information content of the input signal could be preserved while it is passing through the nonlinear device, as we will show under Deterministic Amplitude Resonance with High-Frequency Periodic Carrier-Waves. However, even in this case there is an optimal (range) for the mean-square amplitude of the carrier-wave. Thus, we call this deterministic phenomenon "deterministic amplitude resonance," (DAR), which is also an amplitude-resonance where a deterministic (instead of stochastic/noise) carrier-wave with sufficiently large amplitude produces the optimal signal transfer via the system. See Table 1 for definitions of key terms.

Deterministic Amplitude Resonance with High-Frequency Periodic Carrier-Waves

First, Landa and McClintock (35) realized that SR-like phenomena could occur with high-frequency sinusoidal signals instead of noise. They successfully demonstrated their idea through computer simulations of a binary (double-well potential) SR system. Recently, Mori et al. (36) used high-frequency, noise-free, periodic neural spikes for excitation in a neural computer model to show that SRlike features on the mutual information can be achieved by tuning the frequency of these periodic excitations in the 80–120 Hz range.

In the following sections, we show that high-frequency triangle waves can offer a noise-free signal transfer that can be exactly linear at certain conditions. Sinusoidal waves are also discussed briefly.

The case of triangle (or sawtooth) carrier waves, instead of noise.

Earlier, in a public debate about the future of SR, one of us proposed a noise-free method by using high-frequency triangle waves to improve signal transmission through threshold devices (37) and to reach exactly linear transfer and infinite SNR at the output. Here, we summarize those arguments.

Table 1. Glossary of key terms

Figure 1 shows an example of stochastic resonator hardware with an additive triangle wave, as the carrier-wave, instead of noise. The same argumentation works for sawtooth waves too. Note: the original threshold-based stochastic resonators (17, 18) contain the same hardware elements where Gaussian random noise is used instead of the triangle wave. Due to the binary nature of the visual detection experiments described in this paper, our focus is on subthreshold binary (square-wave) signals with some additional comments about the case of analog signals.

The TE is either an LCD or a comparator. Suppose that the stable output of the LCD is zero and it produces a short uniform positive spike with height U_{LCD} and duration τ whenever the input level crosses the threshold in upward direction. The comparator's output stays at a fixed positive value whenever the input level is greater than the threshold and stays at a lower value (zero or negative) otherwise. Suppose when the input level is greater than the threshold, U_{th} , the comparator output voltage $U_{\text{c}} = U_{\text{H}}$ and otherwise it is 0. The low-pass filter creates a short-time moving average to smooth out the high-frequency components (frequency components due to switching triggered by the carrier wave) and it keeps only the low-frequency part which is the bandwidth of the signal. The parameters, such as the frequency f_s of the signal, the frequency f_t of the triangle wave, and the cut-off frequency f_c of the lowpass filter should satisfy

$$
f_s \ll f_c \ll f_t \ll \frac{1}{\tau} \tag{1}
$$

to transfer the signal with the highest fidelity.

The upper part of Fig. 1 shows the situation without carrier wave: the subthreshold binary signal is unable to trigger the TE thus the output signal is steadily zero. The lower part of Fig. 1 shows the situations where an additive, triangle wave assists the signal to reach the threshold thus it carries the binary signal over the TE resulting in a nonzero output signal. The triangular wave will have to be of a high enough frequency for two main reasons 1) because of the Nyquist sampling theorem, the sampling frequency needs to be at least twice as large as the highest frequency component of the signal, 2) for the low-pass filter to be able to smooth out the carrier signal (which is trash), the carrier signal frequency must be much larger than the reciprocal of the time duration of the binary signal.

The case of level crossing detector. If a constant input signal plus a triangle wave can cross the threshold, the LCD produces a periodic spike sequence with the frequency of the triangle wave. In this situation, the time average of this sequence is $f_t \tau U_{\text{LCD}}$, therefore, for the binary input signal,

Figure 1. Deterministic transfer of subthreshold binary signal through simple threshold-based stochastic resonators with a threshold element [TE: either a level crossing detector (LCD) or a comparator] and an additive triangle wave at the input. Note: the classical threshold-based stochastic resonators contain the same hardware elements except the triangle wave that is substituted by a Gaussian random noise. The role of the low-pass filter is to reduce the amount of irrelevant high-frequency products created by the carrier wave. If those irrelevant high-frequency products are not disturbing, the lowpass filter can be omitted. Top: the subthreshold binary signal is unable to trigger the TE thus the output signal is steadily zero. Bottom: an additive, triangle wave (carrier-wave) assists the signal to reach the threshold thus it carries the binary signal over the TE. The low-pass filter takes a short time average to smooth out the high-frequency components. For high-fidelity transfer, to avoid problems caused by delays or phase shifts, the frequency of the carrier-wave must be much greater than that of the binary signal. In the old stochastic resonance schemes, the carrier-wave was a noise that caused a nondeterministic component (noise) and finite signal-to-noise-ratio (SNR) at the output. The new system is purely deterministic, and its SNR is infinite. Moreover, if the signal is "analog" (continuum amplitude values), the triangle wave with comparator as TE guarantees a linear transfer of the signal provided the threshold level is between the minimum and the maximum of the sum of the signal and the carrier-wave, see in The case of comparator. U_{th} , threshold; U_s , signal; U_t , noise; U_{ICD} , LCD output signal; U_c , comparator output signal; U_{IPF} , signal after low-pass filtering.

the output of the low-pass filtering (LPF) will be binary with amplitude values:

$$
U_{\text{LPF}}(t) = f_{\text{t}} \tau U_{\text{LCD}} \quad \text{or} \quad 0. \tag{2}
$$

Thus, the binary input signal is restored at the output of the LPF without any stochasticity (noise) in it. The only deviation from the input signal is a potentially different amplitude (nonzero amplification) and some softening of the edges dues to the LPF depending on how well Relation 1 is satisfied.

In conclusion, with an LCD as TE, regarding the amplitude resonance versus the carrier wave amplitude U_t , there are three different input amplitude ranges:

- a) $U_s + U_t < U_{th}$ then there is no output signal
- b) $U_{\text{th}} < U_{\text{s}} + U_{\text{t}}$, $U_{\text{s}} < U_{\text{th}}$, $U_{\text{t}} < U_{\text{th}}$ then the binary signal is restored at the output
- c) $U_{\text{th}} < U_{\text{t}}$ then the output is steadily at the high-level $U_{\text{LPF}}(t) = f_{\text{t}} \tau U_{\text{LCD}}.$

Therefore, the binary signal can propagate to the output only in the situation b) when it does that without any noise contribution at the output (the SNR is infinite).

The case of comparator. Note, this system is very different from "Stocks's suprathreshold SR" (19), where a large number of independent comparators with independent noises are used with a common signal and an adder to reach a finite SNR. For the sake of simplicity, but without limiting the generality of the argumentation, suppose that the binary signal, $U_s(t)$, values are 0 and U_s , where $U_s \leq U_{th}$, and the maximum amplitude of the triangle signal, $U_t(t)$ is U_t and its minimum value is 0. In conclusion:

when
$$
U_s(t) + U_t(t) > U_{th}
$$
, $U_c = U_H$ otherwise, $U_c = 0$, (3)

i.e., the comparator's output voltage has only two possible steady states: 0 and U_H . When the input voltage is above the threshold voltage U_{th} the output is U_{H} , otherwise, it is 0.

To evaluate the average output voltage of the comparator, the first question is the fraction of time that the input spends over the threshold, see Fig. 2.

This time T_H within a period of the triangle wave is the period duration $1/f_t$ minus twice the time t_r spent for rising from the minimum to the threshold:

$$
T_{\rm H} = \frac{1}{f_{\rm t}} - 2 t_{\rm r} = \frac{1}{f_{\rm t}} - 2 \frac{U_{\rm th} - U_{\rm s}}{2U_{\rm t}f_{\rm t}} = \frac{U_{\rm t} - U_{\rm th} - U_{\rm s}}{U_{\rm t}f_{\rm t}},\tag{4}
$$

where we used that the slope s of the triangle signal with peak-to-peak amplitude is

$$
s = 2 f_{\rm t} U_{\rm t},\tag{5}
$$

assumed that the signal amplitude U_s is present at the input and assumed Condition 3 that the signal alone is subthreshold, but the sum of the signal and the triangle wave is suprathreshold:

$$
U_{\rm s} + U_{\rm t} \ge U_{\rm th}.\tag{6}
$$

Figure 2. The triangle wave vs. the threshold (U_{th}) .

From Eqs. 3 and 4, the smoothed value of the output voltage $U_{\text{LPF}}(t)$ of the LPF when the input signal amplitude is $U_{\text{S}(t)}=U_{\text{H}}$:

$$
U_{\text{LPF}} = \langle U_{\text{LPF}}(t) \rangle = U_{\text{H}} \frac{T_{\text{H}}}{1/f_{\text{t}}} = U_{\text{H}} \frac{U_{\text{t}} - U_{\text{th}} + U_{\text{s}}}{U_{\text{t}}} = U_{\text{H}} \frac{U_{\text{t}} - U_{\text{th}}}{U_{\text{t}}} + \frac{U_{\text{H}}}{U_{\text{t}}} U_{\text{s}},
$$
(7)

where $\langle \rangle$ denotes short-range averaged (smoothed) value discussed earlier.

The last term on the right side of Eq. 7 demonstrates that the signal amplitude transfers linearly through the system. Therefore, this version of our device is working distortionfree also for analog signals, not only for the present digital signal assumption.

Thus, this device is not only noise-free but also ideally linear for subthreshold signals satisfying Condition 6, even though exact linearity is not an important feature during the experimental study in the present paper.

In conclusion, with a comparator as TE, regarding the amplitude resonance versus the carrier wave amplitude U_t , there are two different input amplitude ranges:

- a) $U_s + U_t < U_{th}$ then there is no output signal
- b) $U_{\text{th}} < U_{\text{s}} + U_{\text{t}}$, $U_{\text{s}} < U_{\text{th}}$, then the binary signal is restored at the output and its amplitude scales inversely with the amplitude U_t of the carrier wave. The maximal amplitude is at $U_t = U_{th}$.

Therefore, the binary signal can propagate to the output only in the b) situation when it does that without any noise (the SNR is infinite) and it has a linear transfer for analog signals. Of note, for a large U_t (triangular waveform) the binary signal is lost as the output would be a constant signal.

The case of sinusoidal carrier waves, instead of triangle waves.

The aforementioned argumentations qualitatively work also for sinusoidal carrier waves except that the linearity of the transfer is lost. The triangle carrier-wave has a Fourier series that has only odd harmonics, where the nth harmonic amplitudes scale with $1/n^2$, that is, the strongest harmonic (the 3rd) is nine times weaker, and the next strongest harmonic (the 5th) is 25 times less than the base harmonic. The qualitative difference is that the absolute value of the slope of the sinusoidal carrier wave is reduced when approaching its peak level and it is zero at the peak. The constant slope of the triangle wave is essential for the exactly linear transfer, see the mathematical proof aforementioned.

In conclusion, when a sinusoidal carrier wave is used instead of a triangle (or sawtooth) wave, the same qualitative features remain, including the zero-noise contribution at the output (infinite SNR). The exception is the linearity of the transfer of analog signals via the comparator that is lost at the sinusoidal carrier wave.

Stochastic Resonance Effects on Neural Processing

In neural systems, it has been demonstrated that responses to externally applied stimuli were maximally enhanced when an optimal level of electrical random noise stimulation was applied. These effects were linked specifically to the opening of voltage-gated sodium $(Na⁺)$ channels in response to electrical stimulation, causing a sodium influx, which in turn causes a local depolarization of the cell membrane (9, 12, 29).

In humans, early SR effects have been mainly demonstrated via behavioral signal detection tasks whereby noise was added to the periphery. For example, the detection of low-contrast visual stimuli was significantly enhanced when the stimuli were superimposed with visual noise (38).

Recently, similar enhancements of visual perception have been reported when noise was directly added to the cerebral cortex by means of transcranial random noise stimulation (tRNS) in studies investigating its acute effects on visual processing (29, 39–44). According to the SR theory, while the optimal level of tRNS benefits performance, excessive noise is detrimental to signal processing (39, 40, 44), resulting in an inverted U-shape relationship between noise benefits and noise intensity. Consistent with SR, tRNS was shown to be particularly beneficial for visual detection performance when the visual stimuli were presented with near-threshold intensity (39, 40, 45).

However, based on the theoretical consideration described earlier, a resonance-like phenomenon can be observed with deterministic stimulations. Here, we test this prediction empirically and investigate if the response of the visual cortex to around-threshold contrast stimuli could also be enhanced via a high-frequency deterministic signal. We tested if triangle or sine waves can modulate signal processing in a resonancelike manner by delivering transcranial alternating current stimulation (tACS) with triangle waveform (tACS_{triangle}) or sine waveform $(tACS_{sine})$ targeting the primary visual cortex (V1) of participants performing a visual contrast sensitivity task and measured their visual detection threshold. We hypothesized that resonance-like DAR effects would be reflected in the beneficial influence of high-frequency stimulation on signal processing via signal enhancement.

MATERIALS AND METHODS

Participants

Individuals with normal or corrected-to-normal vision and without identified contraindications for participation according to established brain stimulation exclusion criteria (46, 47) were recruited in the study. All study participants provided written informed consent before the beginning of each experimental session. Upon study conclusion, participants were debriefed and financially compensated for their time and effort. All research procedures were approved by the Cantonal Ethics Committee Zurich (BASEC Number 2018-01078) and were performed in accordance with the Helsinki Declaration of the World Medical Association (2013 WMA Declaration of Helsinki) and guidelines for noninvasive brain stimulation research through the COVID-19 pandemic (48).

The required sample size was estimated using an a priori power analysis $[G^*Power$ version 3.1 (49)] based on the effect of maximum contrast sensitivity improvement with tRNS shown by Potok et al. (50) ($\eta_p^2 = 0.165$, effect size $f = 0.445$). It revealed that 28 participants should be included in an experirevealed that 28 participants should be included in an experiment to detect an effect with repeated-measures analysis of variance (rmANOVA, 4 levels of stimulation condition), $\alpha =$ 0.05, and 90% power. We included 31 participants in Experiment 1 (tACS_{triangle}) and 32 participants in Experiment 2 (tACSsine) to account for potential dropouts. Visual contrast

detection is potentially prone to floor effects if the contrast detected at baseline approaches the technical limits of the setup. We decided to exclude participants that are exceptionally good in the visual task and present a visual contract threshold below 0.1 (Michelson contrast, see Experimental setup and visual stimuli, for visual contrast intensity range of minimum 0 and maximum 1) in the no tACS baseline condition. For outliers' removal, we used standardized interquartile range (IQR) exclusion criteria (values below Q1–1.5IQR or above $Q3 + 1.5IQR$, where Q1 and Q3 are equal to the first and third quartiles, respectively) to avoid accidental results, unlikely driven by transcranial electrical stimulation (tES), for example, due to participants responding without paying attention to the task.

From the initially recruited sample, we excluded 7 individuals. In tACS_{triangle} Experiment 1: one participant revealed exceptional contrast threshold modulation $(>Q3 + 1.5IQR)$, one participant had a contrast threshold below 0.1 in the baseline condition (also $>Q3 + 1.5IQR$), one participant stopped the session because of unpleasant skin sensations. In $tACS_{sine}$ Experiment 2: one participant revealed exceptional contrast threshold modulation $(>Q3 + 1.5IQR)$, one participant stopped the session because of unpleasant skin sensations, two participants reported frequent (75% accuracy) phosphenes sensation due to stimulation (see tACS characteristics).

The final sample consisted of 28 healthy volunteers (16 females, 12 males; 26.9 ± 4.7 , age range: $21-39$) in tACS_{triangle} Experiment 1, and 28 healthy volunteers (20 females, 8 males; 26.4 ± 4.4 , age range: $20-39$) in tACS_{sine} Experiment 2. We did not collect information about the race of the participants. Twenty of these participants completed both experimental sessions. For participants who took part in both experiments, 15 participants started with tACS_{triangle} and five with tACS_{sine}. The experimental sessions took place on different days with 2.6 ± 1.2 mo on average apart. Delays were caused by the COVID-19 pandemic (48).

General Study Design

To evaluate the influence of tACS on visual contrast detection, we performed two experiments in which we delivered either tACS_{triangle}, or tACS_{sine} targeting V1, during visual task performance (Fig. 3A). In each experiment, three tACS intensities and a control no tACS condition were interleaved in random order. Our main outcome parameter in all experiments was a threshold of visual contrast detection (VCT) that was determined for each of the different tACS conditions (50). The experimental procedure to estimate VCT followed a previously used protocol to assess the influence of tRNS on contrast sensitivity (50). In brief, VCT was estimated twice independently, in two separate blocks within each

Figure 3. Experimental design. A: example trial of 4-alternative forced choice task measuring visual contrast detection threshold (VCT). Transcranial alternating current stimulation (tACS) was delivered for 2 s around the Gabor patch presentation. B: tACS electrodes montage targeting primary visual cortex (V1) and simulation of the induced electric field in the brain. C: example of dose-response psychometric curves and the VCT for the 50% detection accuracy level. We hypothesize that the VCT will be lower (indicating better contrast detection performance of the participant) in one of the tACS conditions (violet) than in the no tACS control condition (blue). D: the order of measurements within each experiment. Each experimental session consisted of application of an anesthetic cream, followed by task training, familiarization protocol, and two independent VCT assessments in 4 interleaved tACS conditions (as specified in A).

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session (Fig. 3D). We determined the individual's optimal tACS intensity (defined as the intensity causing the lowest VCT, i.e., biggest improvement in contrast sensitivity) for each participant in the first block of Experiment 1 (ind $tACS_{triangle}$) and Experiment 2 (ind-t ACS_{sine}) and retested their effects within the same experimental session on VCT data acquired in the second block.

Experimental setup and visual stimuli.

The experiments took place in a dark and quiet room, ensuring similar lighting conditions for all participants. Participants sat comfortably, 0.85 m away from a screen, with their heads supported by a chinrest. Visual stimuli were generated with MATLAB (MATLAB 2020a, The MathWorks, Inc., Natick, MA) using the Psychophysics Toolbox extension that defines the stimulus intensity with Michelson contrast (51–53) and displayed on a CRT computer screen (Sony CPD-G420). The screen was characterized by a resolution of $1,280 \times 1,024$ pixels, a refresh rate of 85 Hz, linearized contrast, and a luminance of 35 cd/m^2 (measured with J17 LumaColor Photometer, Tektronix). The target visual stimuli were presented in the form of the Gabor patch—a pattern of sinusoidal luminance grating displayed within a Gaussian envelope [full width at half maximum of 2.8 cm, i.e., 1° 53' visual angle, with 7.3 cm, i.e., 4° 55' presentation radius from the fixation cross, staying within the central vision, i.e., $\langle 8^\circ \rangle$ radius; (54, 55). The Gabor patch pattern consisted of 16 cycles with one cycle made up of one white and one black bar (grating spatial frequency of 8 c/degree). Stimuli were oriented at 45 tilted to the left from the vertical axis (Fig. 3A), since it was shown that tRNS enhances the detection of low-contrast Gabor patches especially for nonvertical stimuli of high spatial frequency (42).

Four-alternative forced choice visual detection task.

In both experiments, participants performed a visual four-alternative forced choice (4-AFC) visual task, designed to assess an individual VCT, separately for each stimulation condition. A 4-AFC protocol was shown to be more efficient for threshold estimation than commonly used 2-AFC (34). Participants were instructed to fixate their gaze on a cross in the center of the screen. In the middle of each 2.04-s trial, a Gabor patch was randomly presented for 40 ms in one of the eight locations (see Fig. 3A). A stimulus appeared in each location for the same number of times (i.e., 20) within each experimental block in pseudo-randomized order to avoid a spatial detection bias. The possible locations were set on noncardinal axes, as the detection performance for stimuli presented in this way is less affected (i.e., less variable) than when stimuli are positioned on the cardinal axes (56). Each trial was followed by 1-s presentation of only the fixation cross after which the "response screen" appeared. Participants' task was to decide in which quadrant of the screen the visual stimulus appeared and indicate its location on a keyboard (see Fig. 3A). The timing of the response period was self-paced and not limited. Participants completed a short training session (10 trials), with the stimulus presented always at high contrast (0.5; for visual contrast intensity range of minimum 0 and maximum 1), to ensure that they understand the task (see Fig. 3D).

VCT was estimated using the QUEST staircase procedure (32), implemented in the Psychophysics Toolbox in MATLAB

(51–53), which is a method used in psychophysical research to estimate the threshold of a psychometric function (32). The thresholding procedure starts with a presentation of the visual stimulus displayed with 0.5 contrast intensity (Michelson contrast, for visual contrast intensity ranging 0– 1; note that the stimuli were displayed for just 40 ms). When participants answer correctly, QUEST lowers the presented contrast intensity. Consequently, when participants answer incorrectly QUEST increases the presented contrast. The estimated contrast intensity for the next stimulus presentation is based on a maximum-likelihood-based estimate of the underlying psychometric function. Characteristics of the stimuli on each trial are determined by the input stimuli and respective system responses that occurred in the previous sequence of trials (33). The estimated stimulus contrast is adjusted to yield 50% detection accuracy (i.e., detection threshold criterion, Fig. 3C). For a 4-AFC task, 25% accuracy corresponds to a chance level. The remaining parameters used in the QUEST staircase procedure were set as follows: steepness of the psychometric function, $\beta = 3$; fraction of trials on which the observer presses blindly, $\Delta = 0.01$; chance level of response, $\gamma = 0.25$; step size of internal table grain = 0.001; intensity difference between the largest and smallest stimulus intensity, range $=$ 1. VCT was assessed across 40 trials per stimulation condition. Four different conditions were randomly interleaved within each of two experimental blocks (40 trials \times 4 conditions \times 2 blocks; the total number of 320 trials per experimental session, Fig. 3D).

tACS characteristics.

In stimulation trials, tACS (80 Hz) with symmetrical triangle- $(tACS_{triangle})$ or sinewave $(tACS_{sine})$ with no offset was delivered. Stimulation started 20 ms after trial onset and was maintained for 2 s (Fig. 3A). Subsequently a screen with only fixation cross was displayed for 1 s, followed by the selfpaced response time. tACS waveforms were created within MATLAB (MATLAB 2020a, The MathWorks, Inc., Natick, MA) and sent to a battery-driven electrical stimulator (DC-Stimulator PLUS, NeuroConn GmbH, Ilmenau, Germany), operated in REMOTE mode, via a National Instruments I/O device USB-6343 X series, National Instruments). The active tACS conditions and no tACS control condition were interleaved and presented in random order. Timing of the stimuli presentation, remote control of the tACS stimulator, and behavioral data recording were synchronized via MATLAB (MATLAB 2020a, The MathWorks, Inc., Natick, MA) installed on a PC (HP EliteDesk 800 G1) running Windows (Windows 7, Microsoft) as an operating system.

In both experiments, tACS (80 Hz) stimulation (tACS $_{\text{triangle}}$ in Experiment 1 or tACS_{sine} in Experiment 2) was delivered with 0.75 mA, 1 mA, and 1.5 mA amplitude (peak-to-baseline), resulting in maximum current density of 60 $\frac{\mu A}{\text{cm}^2}$, which is below the safety limits of 167 $\frac{\mu A}{\text{cm}^2}$ for transcranial electrical stimulation (57). These intensities were selected based on previous studies investigating the effects of tRNS on contrast sensitivity (39, 50).

Before electrode placement, an anesthetic cream (Emla 5%, Aspen Pharma Schweiz GmbH, Baar, Switzerland) was applied to the position of the intended electrodes on the scalp to numb potential tACS-induced cutaneous sensations and diminish transcutaneous effects of stimulation. To ensure that the cream got properly absorbed, it was left on the scalp for 20 min (58, 59) during which participants completed task training (see Four-alternative forced choice visual detection task and Fig. 3D).

To target V1, we used an electrode montage that was previously shown to be suitable for visual cortex stimulation (39, 50, 60). The electrodes were placed on the head at least 20 min after the application of an anesthetic cream. One tACS 5×5 cm rubber electrode was placed over the occipital region (3 cm above inion, Oz in the 10–20 EEG system) and one 5×7 cm rubber electrode over the vertex (Cz in the 10– 20 EEG system). The electroconductive gel was applied to the contact side of the rubber electrodes (NeuroConn GmbH, Ilmenau, Germany) to reduce skin impedance. The impedance between the electrodes was monitored and kept below 15 k Ω . We used electric field modeling to verify that our electrodes target V1. Simulations were run in SimNIBS 2.1 (61) using the average MNI brain template (see Fig. 3B). Note, that the software enables finite-element modeling of electric field distribution of direct current stimulation without taking into account the temporal characteristics of the alternating current.

Since we used a very brief stimulation time (2 s only), fade-in/out periods were not possible (62). Accordingly, some participants were able to distinguish the stimulation conditions (see RESULTS). We accounted for this possible bias using a control measure and analysis of the potential transcutaneous sensations. In each session, before the start of the main experiment, participants were familiarized with tACS and we assessed the detectability of potential cutaneous sensations (Fig. 3D). The detection task consisted of 20 trials. Participants received either 2 s tACS (0.75, 1, and 1.5 mA tACS $_{\text{triangle}}$ in *Experiment 1* or tACS_{sine} in Experiment 2) or no tACS, to test whether they can distinguish between stimulation versus no stimulation. The task after each trial was to indicate on a keyboard whether they felt a sensation underneath the tACS electrodes. In Experiment 2, an additional control measurement was added to assess the potential phosphene induction by the tACS waveform. tACS (in a lower frequency range) was previously suggested to induce visual phosphenes (63, 64). The protocol was the same with the only difference being that this time after each trial participants indicated on a keyboard whether they perceived any visual sensations while looking on the black computer screen. The determined detection accuracy (hit rates, HR, defined as the proportion of trials in which stimulation is present and the participant correctly responds to it) of the cutaneous sensation (Experiments 1 and 2) and phosphenes (Experiment 2) induced by tACS served as a control to estimate whether any unspecific effects of the stimulation might have confounded the experimental outcomes (62). In the control analysis, we used the HRs for the detection of tACS stimulation conditions (separately for $tACS_{\text{triangle}}$ and $tACS_{\text{sine}}$) as covariates (see Statistical Analysis).

Statistical Analysis

Statistical analyses were performed in IBM SPSS Statistics version 26.0 (IBM Corp.) and JASP (0.16.3) unless otherwise stated. All data were tested for normal distribution using Shapiro–Wilks test of normality. Partial eta-

squared [small $\eta_p^2 = 0.01$, medium $\eta_p^2 = 0.06$, large $\eta_p^2 = 0.14$; (65)] or Cohen's *d* [small *d* – 0.20–0.49 medium *d* – 0.14; (65)] or Cohen's d [small $d = 0.20-0.49$, medium $d =$ 0.50–0.80, large $d > 0.80$; (66)] values are reported as a measure of effect sizes. Standard statistics for simple effects were complemented with their Bayesian equivalents using the Bayes factor (BF_{01}) with $BF_{01} > 1$ indicating evidence in favor of the null hypothesis over the alternative hypothesis. BF_{01} was primarily provided to statistically confirm the lack of an effect throughout the analyses. Variance is reported as SD in the main text and as SE in the figures. Statistical analysis of $tACS_{triangle}$ and $tACS_{sine}$ effects was analogous to the one performed to test highfrequency tRNS (hf-tRNS) effects (50).

Analysis of VCT modulation in tACS $_{\text{triangle}}$ and tACS_{sine} experiments.

First, we tested whether baseline VCT in the no tACS condition differed across the two experimental sessions using a Bayesian independent samples t test (average baseline VCT in blocks 1 and 2 in Experiments 1 and 2) using the BF_{01} .

For all repeated-measures analysis of variance (rmANOVA) models, sphericity was assessed with Mauchly's sphericity test. The threshold for statistical significance was set at $\alpha =$ 0.05. Bonferroni correction for multiple comparisons was applied where appropriate (i.e., post hoc tests; preplanned comparisons of stimulation 0.75 mA, 1 mA, and 1.5 mA vs. no tACS baseline).

To test the influence of tACS_{triangle} on contrast sensitivity, VCT data collected in Experiment 1 (tACS_{triangle}) were analyzed with an rmANOVA with the factors of tACStriangle (no, 0.75 mA, 1 mA, and 1.5 mA $tACS_{triangle}$ and block (1st, 2nd). For each individual and each block, we determined the maximal behavioral improvement, i.e., the lowest VCT measured when tACS_{triangle} was applied, and the associated "optimal" individual $tACS_{triangle}$ intensity (ind- $tACS_{triangle}$). Note, that the ind-tACS_{triangle} was always selected from active stimulation conditions (i.e., even if participants performed better in the no tACS baseline, the ind-tACS intensity was defined based on the lowest VCT during stimulation). The maximal behavioral improvements in the first and the second block were compared using a t test (two-tailed) for dependent measurements. Importantly, we determined ind- t ACS $_{\text{triangle}}$ in the first block, and then used the VCT data of the separate second block to test whether the associated VCT is lower compared with the no tACS condition using t tests for dependent measures. Since we had the directional hypothesis that VCT is lower for the ind-tACS_{triangle} intensity compared with no tACS this test was one-tailed. Determining ind-tACS $_{\text{triangle}}$ and testing its effect on VCT in two separate datasets is important to not overestimate the effect of tACS_{triangle} on visual detection behavior.

Similarly, VCT data collected in *Experiment 2* (tACS_{sine}) was analyzed with a rmANOVA with the factor of tACS_{sine} (no, 0.75 mA, 1 mA, and 1.5 mA tACS $_{\text{sine}}$) and the factor block (1st, 2nd). Again, for each individual and each block, we determined the maximal behavioral improvement and the associated ind-tACS_{sine}. We compared results obtained in the first and second blocks using the same statistical tests as for *Experiment 1*. The maximal behavioral improvements were compared using a t test (two-tailed) for dependent measurements. We examined whether the ind-tACS_{sine},

determined based on the best behavioral performance in first block, caused VCT to be lower compared with the no tACS condition when retested on the independent data set (2nd block) using t tests (one-tailed) for dependent measures.

In both experiments to assess a general modulation of VCT induced by tACS, we calculated the mean change in VCT in all active tACS conditions from first and second blocks normalized to baseline no tACS condition (tACSinduced modulation).

To control for any potential unspecific effects of tACS, we repeated the main analyses of VCT (i.e., rmANOVA) with adding HRs of cutaneous sensation for all current levels (*Experiment 1*, tACS_{triangle} and *Experiment 2*, tACS_{sine}) and phosphene detection (Experiment 2, tACS_{sine}) as covariate. We also tested correlations between the average HR of cutaneous sensation (Experiments 1 and 2) and phosphene (Experiment 2) detection and average tACS-induced modulation using a Pearson correlation coefficient.

Comparison of stimulation-induced VCT modulation in $tACS_{triangle}$, $tACS_{sine}$, and hf-tRNS experiments.

We compared the effects of deterministic transcranial electrical stimulation (tES, i.e., $tACS_{triangle}$ and $tACS_{sine}$) and stochastic tES (i.e., hf-tRNS) on VCT. The data demonstrating the effect of hf-tRNS on VCT were taken from a previous study investigating the effects of hf-tRNS using the same behavioral paradigm (50).

First, we tested whether baseline VCT in the no tES (no tACS, no hf-tRNS) conditions differed across the three experiments using a Bayesian independent samples t test (average baseline VCT in *blocks* $1-2$ in tACS_{triangle}, tACS_{sine}, and hftRNS) using the BF_{01} .

Next, we tested whether a general tES-induced modulation of VCT (mean of all active stimulation conditions from two blocks normalized to baseline no stimulation condition) differed across the three experiments using a Bayesian ANOVA (tES-induced modulation in $tACS_{triangle}$, $tACS_{sine}$, and hf-tRNS experiments) using the BF_{01} .

Finally, we depicted tES-induced modulation of VCT as paired Cohen's d bootstrapped sampling distributions using an online tool [<https://www.estimationstats.com>; (67)]. For each pair of control no tES (i.e., no tACS in tACS $_{\text{triangle}}$, $tACS_{sine}$, and no hf-tRNS) and tES conditions (tACS_{triangle}, tACS_{sine}, and hf-tRNS), a two-sided permutation t tests were conducted. Five thousand bootstrap samples were taken. The confidence interval was bias-corrected and accelerated. The reported P values are the likelihoods of observing the effect sizes if the null hypothesis of zero difference is true. For each permutation P value, 5,000 reshuffles of the control and test labels were performed.

RESULTS

We first tested whether VCT measured during the no tACS conditions differed between the experiments (i.e., average baseline VCT in tACS_{triangle} and tACS_{sine} experiments, see Fig. 4). Bayesian independent samples t test revealed that the baseline VCT measured in the no tACS condition did not differ between experiments ($BF_{01} = 3.439$, i.e., moderate evidence for the H_0).

Figure 4. Average baseline visual contrast detection threshold (VCT) measured in the no transcranial electric stimulation (tES) conditions in transcranial alternating current stimulation (tACS_{triangle}) $n = 28$ (16 females, 12 males), $tACS_{sine}$ n = 28 (20 females, 8 males), high-frequency-transcranial random noise stimulation (tRNS) experiments $n =$ 24 (16 females, 8 males). VCT was assessed for stimuli presented with contrast intensity ranging from 0 to 1. Blue lines indicate mean, gray dots indicate single subject data. Bayes factor (BF_{01}) = 4.869, i.e., moderate evidence for the H_0 .

tACS_{triangle} over V1 Modulates Visual Contrast Threshold

In the first experiment, we investigated whether $tACS_{triangle}$ modulates visual contrast detection when applied to V1. We measured VCT during tACS $_{\text{triangle}}$ at intensities of 0.75, 1, to 1.5 mA peak-to-baseline versus no tACS control condition. We found a general decrease in VCT $(F_{3,81} = 3.41, P = 0.021, \eta_{\rm p}^2 = 0.11$. BE₂₁ – 0.498) reflecting improved contrast sensitivity 0.11, $BF_{01} = 0.498$) reflecting improved contrast sensitivity during tACS $_{\text{triangle}}$ (Fig. 5A). Post hoc comparisons revealed that 0.75 mA and 1 mA stimulation were most effective in boosting contrast processing at a group level, which differed significantly from the no tACS control condition ($P = 0.033$, mean difference, $MD = -6.3 \pm 11.62\%$ and $P = 0.024$, $MD =$ -6.33 ± 10.45 %, respectively). Neither the main effect of block $(F_{1,27} = 2.43, P = 0.13, BF_{01} = 1.429)$ nor tACS_{triangle} \times block interaction ($F_{3,81} = 1.6$, $P = 0.195$) reached significance.

When comparing $tACS_{triangle}$ -induced effects between the first and second block, we found that the maximal behavioral improvement (i.e., maximal tACS_{triangle}-induced lowering of the VCT relative to the no tACS condition) was not significantly different between the first (MD = $-14.64 \pm$ 12.6%, VCT decrease in 25 out of 28 individuals) and the second block ($MD = -15.75 \pm 15.73\%$, VCT decrease in 24 out of 28 individuals; $t_{27} = 0.604$, $P = 0.551$, $BF_{01} = 4.219$), additionally showing that no time effects arose from the first to the second block of measurement.

Next, we defined the optimal ind- $tACS_{triangle}$ for each participant and examined whether its effects can be reproduced. We observed that the ind-tACS $_{\text{triangle}}$ determined in first block (Fig. 5B) caused a decrease in VCT compared with the no tACS condition when retested within the same experimental session ($t_{27} = 1.84$, $P = 0.039$, $BF_{01} = 0.463$, VCT decrease in 18 out of 28 individuals, $MD = -5.26 \pm 18.23$ %, Fig. 5C). Note, that the aforementioned analysis does not contain an element

Figure 5. The effect of transcranial alternating current stimulation (tACS_{triangle}) on visual contrast detection threshold (VCT) measured in *Experiment 1*. VCT was assessed for stimuli presented with contrast intensity ranging from 0 to 1. A: effect of tACS $_{\rm triangle}$ on VCT on a group level measured across first and second blocks. Decrease in VCT reflects improvement of visual contrast sensitivity. VCT in tACStriangle conditions normalized to the no stimulation baseline. All data means ± SE; *P < 0.05, repeated-measures analysis of variance (rmANOVA). B: individually defined optimal tACS_{triangle} based on behavioral performance during the first block. C: detection improvement effects of individualized tACS_{triangle} (selected based on block 1) measured on the independent VCT data of block 2. Gray dots indicate single subject data; *P < 0.05, t test for dependent measures. $n = 28$ (16 females, 12 males).

of intrinsic circularity because the ind-tACStriangle and the VCT measure were based on independent data sets.

The cutaneous sensation control experiment revealed that some of our participants could detect $tACS_{triangle}$ conditions (HR at 0.75 mA = 12.5 ± 25 %, 1 mA = 18.75 ± 27.74 %, 1.5 mA = 41.07 \pm 43.68%, mean HR = 24.11 \pm 27.34%). We reanalyzed our main outcome parameter by adding sensation detection HRs for each current level as covariates (HRs were z-scored because of non-normal distribution). The main effect of tACStriangle remained significant $(F_{3,72} = 3.36, P = 0.023, \eta_p^2 = 0.12)$.
Moreover the mean HR of cutaneous sensation detection did Moreover, the mean HR of cutaneous sensation detection did not correlate with the average tACS_{triangle}-induced VCT modulation ($r = 0.181$, $P = 0.357$, $BF_{01} = 2.835$), making it unlikely that transcutaneous sensation was the main driver of our results.

tACS_{sine} over V1 Modulates Visual Contrast Threshold

In the second experiment, we explored the effects of tACSsine applied over V1 on visual contrast detection. VCT

was measured during tACS_{sine} at intensities of 0.75, 1, to 1.5 mA peak-to-baseline versus no tACS control condition. We observed a general decrease in VCT with increasing tACS_{sine} intensity $(F_{3,81} = 4.78, P = 0.004, \eta_p^2 = 0.15, BF_{01} = 0.111$ reflecting improved contrast sensitivity during 0.111) reflecting improved contrast sensitivity during tACSsine. Post hoc comparisons revealed that the 1 mA and 1.5 mA stimulation were most effective in enhancing contrast processing, which differed significantly from the no tACS control condition ($P = 0.042$, MD = $-8.04 \pm 13.82\%$) and $P = 0.008$, MD = -6.52 \pm 12.66%, respectively, Fig. 6A). There was no main effect of block ($F_{1,27} = 0.02$, $P =$ 0.878, BF₀₁ = 3.619) or tACS_{sine} \times block interaction ($F_{3,81}$ = $0.5, P = 0.684$.

When comparing tACS_{sine}-induced effects between the first and second block, we found that the maximal behavioral improvement, defined as maximal $tACS_{sine}$ -induced lowering of the VCT was not different between the first $(MD = -17.78 \pm 15.82\%, VCT$ decrease in 25 out of 28

Figure 6. The effect of transcranial alternating current stimulation (tACS)_{sine} on visual contrast detection threshold (VCT) measured in Experiment 2. VCT was assessed for stimuli presented with contrast intensity ranging from 0 to 1. A: effect of tACS_{sine} on VCT on a group level measured across first and second blocks. Decrease in VCT reflects improvement of visual contrast sensitivity. VCT in tACS_{sine} conditions normalized to the no stimulation baseline. All data means \pm SE; *P < 0.05, **P < 0.01, repeated-measures analysis of variance (rmANOVA). B: individually defined optimal tACS_{sine} based on behavioral performance during the first block. C: detection improvement effects of individualized tACS_{sine} (selected based on block 1) measured on the independent VCT data of block 2. Gray dots indicate single subject data; $*P < 0.01$, t test for dependent measures. $n = 28$ (20 females, 8 males).

individuals) and the second block (MD = $-18.37 \pm 16.67\%$, VCT decrease in 22 out of 28 individuals; $t_{27} = 0.95$, $P =$ 0.353, $BF_{01} = 3.320$.

We determined the optimal ind-t $\rm{ACS}_{\rm{sine}}$ and tested whether its effects can be reproduced. Similar to ind $tACS_{triangle}$ in Experiment 1, the optimal ind-tACS_{sine} determined in first block (Fig. 6B) significantly lowered the VCT compared with the no tACS condition when retested on the independent VCT data set of the second block ($t_{27} = 2.59$, $P = 0.008$, $BF_{01} = 0.157$, VCT decrease in 18 out of 28 individuals, $MD = -7.85 \pm 21.84\%$, Fig. 6C).

Similarly to *Experiment 1*, we assessed the HR of cutaneous sensation detection (HR at 0.75 mA = $16.07 \pm 27.4\%$, $1 \text{ mA} = 21.43 \pm 30.21\%, 1.5 \text{ mA} = 50.89 \pm 43.29\%, \text{mean HR} =$ 29.46 ± 27.36%). We reanalyzed our main outcome parameter by adding mean cutaneous sensation detection HRs as a covariate (HRs were z-scored because of non-normal distribution). The main effect of $tACS_{sine}$ remained significant $(F_{3,72} = 4.67, P = 0.005, \eta_p^2 = 0.16)$. The mean HR of cutaneous sensation did not correlate with the average cutaneous sensation did not correlate with the average tACS_{sine}-induced VCT modulation ($r = -0.12$, $P = 0.542$, $BF_{01} = 3.569$). In this experiment, we additionally tested phosphenes detection (HR_{phos} at 0.75 mA = $3.57 \pm 8.91\%$, $1 \text{ mA} = 5.36 \pm 12.47\%, 1.5 \text{ mA} = 6.25 \pm 16.14\%, \text{ mean HR} =$ 5.06 \pm 10.48%). After adding HR_{phos} as covariate (*z*-scored HR_{phos}), the main effect of tACS_{sine} remained significant $(F_{3,72} = 4.82, P = 0.004, \eta_p^2 = 0.17)$. Accordingly, the mean HR of phosphene detection did not correlate with the average tACS_{sine}-induced VCT modulation ($r = -0.14$, P = 0.493, $BF_{01} = 3.405$.

Comparison of $tACS_{triangle}$, tACS $_{sine}$, and hf-tRNS-Induced Modulation

First, we tested whether baseline VCT measured during the no tES conditions differed between the experiments (i.e., average baseline VCT in tACStriangle, tACSsine, and hf-tRNS experiments, see Fig. 4). Bayesian independent samples t test revealed that the baseline VCT measured in the no tES condition did not differ between experiments ($BF_{01} = 4.869$, i.e., moderate evidence for the H_0).

Next, we compared tES-induced modulation effects between experiments ($tACS_{triangle}$, $tACS_{sine}$, and hf-tRNS experiments, see Fig. 7A). A Bayesian ANOVA revealed that the general tES-induced modulation did not differ between experiments ($BF_{01} = 8.956$, i.e., moderate evidence for the H₀), suggesting that all three stimulation types were equally effective in lowering VCT.

Finally, we assessed the strength of the tES-induced effects on VCT across tACStriangle, tACSsine, and hf-tRNS experiments defined as paired Cohen's d bootstrapped sampling distributions (see Fig. 7B). We found comparable (small) effects of significant differences between no tES baseline VCT and averaged VCT in active tES conditions in all experiments using the two-sided permutation t test [in tACS_{triangle} $d = -0.17$ (95.0% CI -0.284; -0.0698) P = 0.0034; in tACS_{sine} $d = -0.242$ (95.0% CI -0.444; -0.103), $P = 0.0016$; in hf-tRNS $d = -0.249$ (95.0% CI -0.433; -0.088) $P = 0.0092$. The effect sizes and confidence intervals (CIs) are reported above as: effect size (CI width lower bound; upper bound).

Figure 7. Comparison of transcranial alternating current stimulation (tACS)_{triangle} ($n = 28$; 16 females, 12 males), $tACS_{sine}$ ($n = 28$; 20 females, 8 males), and hf-transcranial random noise stimulation (tRNS)-induced modulation ($n =$ 24; 16 females, 8 males). A: VCT modulation induced by tACS_{triangle}, tACS_{sine}, hf-tRNS. The general modulation of VCT induced by transcranial electric stimulation (tES) was calculated as mean of all active tES conditions from first and second blocks normalized to baseline no tES condition in each experiment. Decrease in VCT reflects improvement of visual contrast sensitivity. $*P < 0.01$, two-sided permutation t test. B : the paired Cohen's d for three comparisons shown in the Cumming estimation plot. Each paired mean difference is plotted as a bootstrap sampling distribution. Mean differences are depicted as dots, 95% confidence intervals are indicated by the ends of the vertical error bars.

DISCUSSION

Theoretical modeling shows that adding a deterministic, high-frequency sinusoidal signal instead of stochastic noise could lead to signal enhancement due to resonance, according to the DAR mechanism. Our experimental proof-of-concept study revealed that stimulation of V1 with a deterministic tACS signal instead of stochastic noise leads to signal enhancement in visual processing. We measured visual contrast sensitivity during $tACS_{triangle}$ and $tACS_{sine}$. On the group level, we found consistent tACS_{triangle}- and tACS_{sine}-induced decrease in VCT, reflecting enhancement in visual contrast processing during V1 stimulation (Fig. 5A and Fig. 6A). The online modulation effects of individually optimized $tACS_{triangle}$ and tACS_{sine} intensities (Fig. 5B and Fig. 6B) were replicated on the independent VCT data (Fig. 5C and Fig. 6C). Finally, we demonstrated that the effects of deterministic stimulation on VCT are comparable with stochastic stimulation of V1 with hf-tRNS (Fig. 7, A and B).

tACS with Triangle and Sine Waveform Improve Visual **Sensitivity**

Our findings provide the first proof of concept that the deterministic $tACS_{triangle}$ and $tACS_{sine}$ delivered to V1 can modulate visual contrast sensitivity. Across two experiments we showed that the modulatory effects of tACS on visual sensitivity are not waveform specific, as both $tACS_{triangle}$ and tACS_{sine} induced significant decrease in VCT (Fig. 5A and Fig. 6A).

One of the main characteristics of SR-like effects is the optimal intensity of noise, which is required to yield improved performance (5, 15). Here, we did not observe an excessive level of tACS that would be detrimental to visual processing (Fig. 5A and Fig. 6A). This is consistent with our predictions that in line with DAR [see mathematical predictions in The case of triangle (or sawtooth) carrier-waves, instead of noise] adding high-frequency deterministic signal should result in a noise-free output where the detection processing is not disturbed by random stimulation effects.

Similar to other studies investigating resonance-like effects (39, 50), our results have revealed large variability among participants in terms of the optimal intensity resulting in the strongest modulation of visual contrast sensitivity (Fig. 5B and Fig. 6B). However, consistent with the effects of tRNS-induced online modulation of contrast processing in V1 shown previously (39, 50), the effects of individualized tACS intensity were replicated on the independent VCT data set collected within the same experimental session (Fig. 5C and Fig. 6C), suggesting consistent beneficial influence of tACS on signal enhancement.

We implemented several control measures to test whether the improvement in visual processing was driven by effective stimulation of V1 rather than any unspecific effects of tACS. We applied an anesthetic cream to numb potential stimulation-induced cutaneous sensation on the scalp (58, 59). Although the anesthetic cream numbs the skin and reduces the cutaneous sensations resulting from tACS, it does not eliminate them completely in all individuals. The control cutaneous sensation detection assessment in the current study showed that some participants could accurately detect tACS, and that the mean detection rate was rather low

(mean HR = 24.11 \pm 27.34% in tACS_{triangle} and mean HR = $29.46 \pm 27.36\%$ in tACS_{sine}). Cutaneous sensation and phosphenes detection (also very low, mean $HR_{\text{phos}} = 5.06 \pm$ 10.48%) did not correlate with the average tACS-induced VCT modulation neither in tACStriangle, nor in tACSsine experiment. Moreover, stimulation effects remained significant in the additional analysis using tactile or phosphene sensation detection during $tACS_{triangle}$ and $tACS_{since}$ as covariate.

Although $tACS_{sine}$ is a well-established and frequently used noninvasive brain stimulation method, high-frequency $tACS_{sine}$ is less common. The effects of 80-Hz $tACS_{sine}$ were sporadically tested in the past using physiological and behavioral paradigms. Ten minutes of 140-Hz tACS_{sine} was shown to increase primary motor cortex (M1) excitability as measured by transcranial magnetic stimulation-elicited motor-evoked potentials during and for up to 1 h after stimulation. Control experiments with sham and 80-Hz stimulation did not show any effect, and 250-Hz stimulation was less efficient, with a delayed excitability induction and reduced duration (68). The researchers postulated that the changes in corticospinal excitability result from externally applied high-frequency oscillation in the ripple range (140 Hz corresponding to middle, 80 Hz lower, and 250 Hz upper border) that interfere with ongoing oscillations and neuronal activity in the brain (68). We can, however, not directly translate the effects of tACS_{sine} of M1 to our stimulation of V1. In addition, the stimulation effects observed in our study are likely reflecting acute modulation of contrast processing, as stimulation was only applied for short intervals (2 s) always interleaved with control (no tACS) condition. Thus, it is possible that even though 80-Hz stimulation did not lead to long-term effects on cortical excitability it can still affect cortical processes acutely.

In the visual domain, 1.5 -mA high-frequency $tACS_{sine}$ was applied to V1 for 15–45 min in a study investigating the effect of covert spatial attention on contrast sensitivity and contrast discrimination (69). That study found that contrast discrimination thresholds decreased significantly during 60-Hz $tACS_{sine}$, but not during 40- and 80-Hz stimulation. This previous study used, however, different visual stimuli than that utilized here, i.e., a random dot pattern. Moreover, they used a more complicated behavioral paradigm, where contrastdiscrimination thresholds were tested using two attention conditions, i.e., with or without a peripheral cue, as the study goal was to explore the influence of attentional processes on visual tasks. One tACS mechanism that has been tested in the visual domain is the reduction of adaptation (70, 71). More specifically, a seminal study has shown that 10-Hz tACS reduces sensory adaptation in a visual motion perception task (70). Since sensory adaptation increases thresholds for detection, potentially reducing adaptation during tACS could result in decreased thresholds. However, the aforementioned study design differed substantially from the one presented here. Kar and Krekelberg (70) used a 40-s adaptor stimulus to induce adaptation while our stimuli were presented for only 40 ms (i.e., 3 magnitudes shorter). This is relevant because it was previously shown that adaptation gets stronger and lasts longer with increasing adaptation duration (72). Moreover, Kar and Krekelberg (70) used 10-Hz tACS that was substantially lower than the one exploited in our

experiments (i.e., 80 Hz), making direct comparisons difficult. Overall, additional experiments would be required to test whether our results could be explained by a reduction of visual adaptation.

Even though the vast majority of tACS studies to date have used a sinusoidal waveform, an alternating current does not have to be sinusoidal, since it can take any arbitrary waveform such as rectangular wave (73), pulsed (74), or sawtooth (75). Dowsett and Herrmann (75) investigated the effects of sinusoidal and sawtooth wave tACS on individual endogenous α -power enhancement. They observed α oscillations enhancement both during and after sawtooth stimulation. The effect seemed to depend on the shape of the sawtooth, as they found that positive, but not negative, ramp sawtooth significantly enhanced α power during stimulation relative to sham. They postulated that a sudden, instantaneous change in current might be more effective than a sinusoidal current in increasing the probability of neurons firing. In this regard, Fröhlich and McCormick (Supplementary Material in Ref. 76) demonstrated that ramps of increasing voltage with a steeper gradient resulted in increased neural firing in vitro, relative to ramps with a low gradient but reaching the same maximum voltage. This suggests that it is not only the total amount of current but also the rate of change of current that can modulate neural firing. Note, that triangle waveform has a faster rate of change of current than the sine wave.

Although we postulate that the effect of tACS on VCT in our study results from a resonance-like mechanism, this is not the only potential mechanism. Importantly, the commonly accepted mechanism of action of tACS is that it entrains action potential firing, and thus neural oscillations (27). The entrainment effect anticipates a monotonic relationship between the tACS effect and intensity, where increasing stimulation intensity results in greater effects for stimulation waveforms that are tuned to the endogenous oscillation (77). The effects of tACS in regard to induced brain oscillations seem to depend on the stimulation duration (78). Although the entrainment after-effects were observed after tACS had been delivered for several minutes (78), a short stimulation of 1 s did not produce after-effects on the amplitude or phase of the electroencephalogram (79). Moreover, in the study investigating the effects of intermittent α tACS of either 3 s or 8 s, the after-effects were found only for the 8-s condition (80). The authors excluded entrainment as a potential underlying mechanism and postulated plasticityrelated changes as the responsible mechanism for the observed after-effects. Here, we used very brief stimulation of 2-s tACS per trial, a duration seemingly too short to induce the entrainment effects on cortical processing.

Furthermore, it was postulated that a very small amount of applied electric field can bias spike timing or spike probability when a neuron nears the threshold of spike generation (81). Accordingly, it was shown that although entrainment effects can arise at field strengths <0.5 mV/mm, physiological effects are more pronounced for higher intensities (around 1 mV/mm), according to intracranial recordings in awake nonhuman primates (82). These values are well above the simulated-induced electric field in our study (around 0.2 mV/mm, see Fig. 3). Further studies are required to fully disentangle the underlying neuronal effects of tACS driving the

enhancement in visual detection. To exclude the influence of entrainment on VCT modulation a jittered tACS protocol could be used. A paradigm using stimulation of jittered flickering light, where instead of a rhythmic flicker, interstimulus intervals of the square wave were jittered with a maximum of ±60%, was shown to fail in inducing rhythmic brain response (83). If a jittered tACS of V1 would still influence contrast sensitivity we could assume the nonentrainment origin of the effect.

Comparison of tACS_{triangle}, tACS_{sine}, and hf-tRNS

In the phenomenon of SR, random noise added to a nonlinear system can increase its responsiveness toward weak subthreshold stimuli. One aim of the present study was to explore whether a deterministic and periodic signal can substitute stochastic noise and still lead to response enhancement in a threshold-based stochastic resonator. The DAR characteristics of high-frequency deterministic signals might offer a noise-free output, thus additionally increasing SNR. We proposed the following testable hypotheses: 1) tACS_{triangle} will have a larger resonance-like effect compared with hf $tRNS$, 2) tACS_{sine} will have less effect than tACS_{triangle}, due to the loss of waveform linearity. We found enhancement effects of both $tACS_{triangle}$ and $tACS_{sine}$ (Fig. 5A and Fig. 6A), however, to test whether these effects are indeed superior to stochastic stimulation, we directly compared the VCT modulation induced by tACS_{triangle}, tACS_{sine}, and hf-tRNS (Fig. 7). The baseline contrast sensitivity between the compared experiments was not different (Fig. 4). Counter to our hypothesis, the noise-free tACS did not result in stronger contrast sensitivity enhancement, as average VCT modulation did not differ between the three stimulation conditions, as confirmed by Bayesian analysis (Fig. 7A). Accordingly, the effects sizes of all three stimulation types were comparable (Fig. 7B). Therefore, we showed that both deterministic and stochastic high-frequency stimulations were equally effective in inducing resonance-like effects.

In real life (compared with mathematical simulations) neural processing is intrinsically noisy. How this intrinsic noise interacts with the applied noise/SR signal will have implications for the validity of our mathematical model. The task we used is a 4-AFC discrimination task, which means that when tRNS is added to the neurons then there must be a distinction between three noisy locations and one signal and noise location. How added noise influences this comparison remains unresolved and an area for further investigation. To investigate this, a two-sided tRNS experiment could be run where noise is added to the left and/or right V1 (or S1) to explore the influence of more noise added to the system where the signal is not present versus when it is present. This might give some information on how the brain interacts with signals, added noise, and intrinsic noise compared with intrinsic noise only on a discrimination task.

The optical excitations in this work were square wave signals (i.e., the visual stimulus was switched on for 40 ms). Open theoretical and related experimental questions are: What is the role of the finite time it takes for the square wave signal to reach its amplitude? What if the signal has a

different periodicity? Possible interrelations between the duration of the pulses and the interpulse intervals?

Conclusions

The present study provides the first evidence for resonance-like neural signal enhancement without adding a stochastic noise component. We showed that "deterministic" 80-Hz tACS and "stochastic" hf-tRNS are equally effective in enhancing visual contrast detection. In the range of commonly used intensities of tES to induce SR, tACS did not result in detrimental effects related to the excessive interference signal, thus providing increased SNR in all tested intensities, according to DAR predictions. These findings shed new light on the effects induced by both 80-Hz tACS and hftRNS, and their underlying mechanisms.

DATA AVAILABILITY

Data will be made available upon reasonable request.

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DISCLOSURES

No conflicts of interest, financial or otherwise, are declared by the authors.

AUTHOR CONTRIBUTIONS

W.P., O.v.d.G., S.S., M.B., L.B.K., and N.W. conceived and designed research; W.P. and S.S. performed experiments; W.P. and S.S. analyzed data; W.P., O.v.d.G., S.S., L.B.K., and N.W. interpreted results of experiments; W.P. prepared figures; W.P. drafted manuscript; W.P., O.v.d.G., F.F., L.B.K., and N.W. edited and revised manuscript; W.P., O.v.d.G., S.S., M.B., F.F., L.B.K., and N.W. approved final version of manuscript.

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