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A Systematic Review and Large-Scale tES and TMS Electric Field Modeling Study Reveals How Outcome Measure Selection Alters Results in a Person- and Montage-Specific Manner

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Abstract

Background. Electric field (E-field) modeling is a potent tool to examine the cortical effects of transcranial magnetic and electrical stimulation (TMS and tES, respectively) and to address the high variability in efficacy observed in the literature. However, outcome measures used to report E-field magnitude vary considerably and have not yet been compared in detail.

Objectives. The goal of this two-part study, encompassing a systematic review and modeling experiment, was to provide an overview of the different outcome measures used to report the magnitude of tES and TMS E-fields, and to conduct a direct comparison of these measures across different stimulation montages.

Methods. Three electronic databases were searched for tES and/or TMS studies reporting Efield magnitude. We extracted and discussed outcome measures in studies meeting the inclusion criteria. Additionally, outcome measures were compared via models of four common tES and two TMS modalities in 100 healthy younger adults.

Results. In the systematic review, we included 118 studies using 151 outcome measures related to E-field magnitude. Structural and spherical regions of interest (ROI) analyses and percentilebased whole-brain analyses were used most often. In the modeling analyses, we found that there was an average of only 6% overlap between ROI and percentile-based whole-brain analyses in the investigated volumes within the same person. The overlap between ROI and whole-brain percentiles was montage- and person-specific, with more focal montages such as 4x1 and APPS-tES, and figure-of-eight TMS showing up to 73%, 60%, and 52% overlap between ROI and percentile approaches respectively. However, even in these cases, 27% or more of the analyzed volume still differed between outcome measures in every analyses.

Conclusions. The choice of outcome measures meaningfully alters the interpretation of tES and TMS E-field models. Well-considered outcome measure selection is imperative for accurate interpretation of results, valid between-study comparisons, and depends on stimulation focality and study goals. We formulated four recommendations to increase the quality and rigor of E-field modeling outcome measures. With these data and recommendations, we hope to guide future studies towards informed outcome measure selection, and improve the comparability of studies.

1. Introduction

Electric field (E-field) modeling is a computational approach to estimate the amount of transcranial electrical stimulation (tES) and transcranial magnetic stimulation (TMS) that reaches the cortex [1, 2]. By segmenting an individual's structural magnetic resonance imaging (MRI) scan into different tissue types such as skin, bone, cerebrospinal fluid (CSF), grey matter, and white matter, it is possible to simulate the magnitude of E-fields induced by tES and TMS. By doing so, E-field modeling provides a potent tool to individually examine the effects of noninvasive brain stimulation and to address the high variability in efficacy that is currently observed across individuals. In the past, E-field modeling has already helped researchers derive novel tES montages and identify dose-response relationships [3-8], has suggested optimal stimulation targets in clinical cohorts [9, 10], and has pinpointed which cortical regions are being stimulated by TMS [11]. In recent years, the introduction of software packages such as SimNIBS and ROAST has catalyzed the widespread use of E-field modeling [1, 2]. However, despite the standardization offered by these software packages, crucial experimental decisions made by researchers can vary considerably between studies, affecting study results [12-16]. In particular, an often-overlooked determinant of E-field modeling findings is the selected outcome measure to quantify E-field magnitude, which may affect interpretation.

To date, researchers have used numerous outcome measures to quantify the magnitude of E-fields produced by tES and TMS. For instance, E-field magnitude has been quantified both in a region of interest (ROI) and as a percentile of the total induced E-field magnitude. While both metrics pursue a common goal, they do so in a substantially different manner. Whereas the ROI focuses on a specific, user-defined brain region (e.g., the left primary motor cortex [M1]) [17-22], a percentile whole-brain approach studies E-fields across the entire brain and does not restrict the analyses to any region [23-25]. Moreover, even within the same approach, wide methodological variations prevail. Some studies have quantified E-field magnitude within spherical ROIs of considerably different radii, ranging between 0.5 and 45 mm, whereas others used different geometric shapes such as a cubical ROI [17-22].

While numerous studies have examined the importance of methodological factors such as MRI parameters [12, 14, 26, 27], head model detail [11, 27, 28] and meshing approach [2, 29], systematic investigation of the impact of modeling outcome measure choice on E-field magnitude results has not yet been achieved. Critically, the choice of outcome measure affects all modeling approaches even as more advanced segmentation and meshing techniques emerge, making it a key consideration both now and into the future. As focus shifts towards unraveling dose-response curves associated with E-field magnitude [30-34], it is crucial that we are comparing the same brain regions and volumes necessitating a critical evaluation of outcome measures.

Therefore, this two-part study set out to formally examine and compare the breadth and frequency of different outcome measures that have previously been used to quantify E-field magnitude. In Part 1, we conducted a systematic review of the literature to identify and describe all the different outcome measures that researchers have used to quantify tES and TMS induced E-field magnitude. While a recent narrative review reported the use of E-field modeling within the scope of a specific software package [35], we sought to systematically report which outcome measures have been used so far across all software packages and populations. In Part 2, to further facilitate the interpretation of previous work and inform future work, we used E-field modeling and examined the most common approaches extracted from the systematic review in a large open-source dataset of 100 healthy adults to elucidate the impact of outcome measure choice. As different modalities stimulate varying volumes of brain tissue and focus E-fields in different ways, we simulated and compared outcome measures across 4 tES and 2 TMS montages (600 total models).

2. Methods

2.1. Systematic Review: Eligibility, Search Strategy and Extracted Information

This review was conducted according to the Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) Statement [36]. We consulted three electronic databases (PubMed, Scopus and Web of Science) to examine how E-field magnitude is quantified. We included studies if they adhered to the following eligibility criteria: (1) full-text availability; (2) written in English; (3) modeling of tES and/or TMS in humans; (4) reporting E-field magnitude. Studies were excluded if insufficient information was provided to reproduce the E-field magnitude extraction procedure. Given the significant advances made in the modeling field in the past decade, we confined our literature search to 2012–2022, with the final search taking place on December 5th, 2022. Our search keys are shown in **Supplementary materials 1**. We extracted the approach used to quantify E-field magnitude from each included study. As we did not extract the outcome measures of the included studies, but rather the approach to obtain E-field magnitude, we did not perform risk of bias analyses.

2.2. Computational modeling

2.2.1. Head Model Creation Overview

All simulations were performed in SimNIBS v4.0.0. To disentangle how different outcome measures affect E-field magnitude, we retrieved T1w and T2w structural MRI scans from 100 participants from the Human Connectome Project dataset (22–35 years, 50 females) [37]. Through the SimNIBS – Charm pipeline, these MRI scans were segmented and meshed into tetrahedral head models with 10 compartments (air, eyes, skin, muscle, compact bone, spongy bone, cerebrospinal fluid, veins, grey matter, and white matter; **Figure 1**) [1, 29]. All head models were visually inspected and we confirmed that there were no segmentation errors present.

2.2.2. Electric Field Models

Four common tES and two common TMS montages targeting left M1 were modeled. Standard conductivity values were used (**Supplementary materials 2**). We performed all tES simulations at an intensity of 1 mA, and all TMS simulations at 50% stimulator output on a MagPro R30 machine (dl/dt = 75×10^6 A/s). As the induced E-field magnitude is linearly proportional to the stimulation intensity, multiplying and dividing the obtained E-field strengths is a simple heuristic to convert the 1 mA and 50% stimulator output results to other intensities.

For tES, we simulated the following four montages (**Figure 1**). (1) APPS-tES consisting of two rectangular 1 by 1 cm electrodes, with the anode placed over CP3 and the cathode over FC3 [4]. (2) Bilateral tES consisting of two rectangular 5 by 5 cm electrodes, with the anode placed over C3 and the cathode over C4 [38]. (3) M1-SO tES consisting of two rectangular 5 by 5 cm electrodes, with the anode placed over C3 and the cathode over Fp2 [39]. (4) 4x1 tES consisting of 5 circular electrodes (r = 0.25 cm), with the anode placed over C3 and the four cathodes placed over FC3, C1, CP3 and C5 [3]. Here, the current going through each cathode was adjusted to -0.25 mA, to let 1 mA flow through the anode.

For TMS, we modeled two coils placed over C3 with a 45° angle to the midline (**Figure 1**). (1) Figure-of-eight TMS consisted of a MagVenture MC-B70 coil [40]. (2) Circular TMS consisted of a MagVenture MMC-140-II coil [40].

After a formal assessment (**Supplementary materials 3**), we replaced values exceeding the 99.9th percentile E-field magnitude with the 99.9th percentile E-field magnitude. This was done as higher values are prone to staircasing errors, and lower values underestimate the peak magnitude [41].

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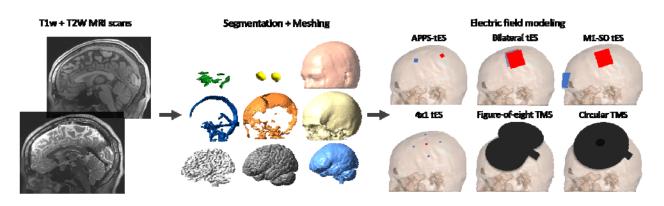


Figure 1. Electric field modeling pipeline. Per participant, we simulated four transcranial electrical stimulation (tES) and two transcranial magnetic stimulation (TMS) modalities.

2.2.3. Qualitative and Quantitative Assessments of Outcome Measures

Based on the systematic review (Cf., **3.1. Systematic Review Results**), we discussed the spherical, structural, and cubical ROI and whole-brain percentile approaches. Next to a qualitative discussion, all outcome measures except for the structural ROI were subjected to quantitative analyses. The structural ROI was not included in quantitative analyses as it is highly specific, based on the individual's neuroanatomy, neurophysiology, and/or the atlas used to define it. Therefore, quantifying this method has limited applicability and generalizability. The spherical and cubical ROIs were centered at the subject-space transformed cortical projection of C3 [42].

For the spherical ROI, we examined how ROI size affects E-field magnitude by extracting the mean E-field magnitude obtained from ROIs with radii ranging between 0.5 and 45 mm. These radii were selected as outer limits based on our systematic review [21, 22]. We also extracted the percentile E-field magnitude from within these ROIs, with percentiles ranging from the 10th to 99.9th percentile. This range was selected to ensure a complete overview, with the 25th and 100th percentile being the lowest and highest percentiles used in the studies included in the systematic review [43, 44]. Finally, we correlated mean E-field magnitudes

obtained from ROIs with differing sizes against each other per modality through Spearman's correlations.

For the cubical ROI, we construed ROIs with identical volumes as the spherical ROIs to assess whether both approaches result in similar mean E-field magnitudes.

For the percentile-based whole-brain analyses, we extracted the 10th to 99.9th percentile E-field magnitude, the analyzed grey matter volume per percentile (mm³) and the Spearman's correlation between different percentiles within the same montage.

Due to the computational nature of E-field models, no inferential statistical analyses were conducted.

2.3. Comparison of Different Outcome Measures

In addition to separate ROI and whole-brain percentile analyses, we directly compared common outcome measures per montage to aid interpretation of prior literature and guide future study methodologies. Based on the systematic review (Cf., **3.1. Systematic Review Results**), we compared the spherical ROI approach (radii ranging between 0.5 and 45 mm) against the percentile-based whole brain approach (percentiles ranging between the 10th and 99.9th percentile). In each modality and radius – percentile combination, we calculated the difference and Spearman's correlation between E-field magnitudes obtained by both approaches, and the volumetric overlap. Volumetric overlap was quantified as the grey matter volume (mm³) included in both approaches, divided by the grey matter volume included by at least one approach.

3. Results

3.1. Systematic Review Results

We included 118 studies after removal of duplicates and title, abstract, and full-text screening. In these studies, E-field magnitude was reported 151 times as several studies reported more than one outcome measure (**Figure 2** and **Table 1**). Studies either extracted E-

field magnitude from within a ROI (n=100, 71/29 tES/TMS) or the whole-brain (n=51, 16/35 tES/TMS).

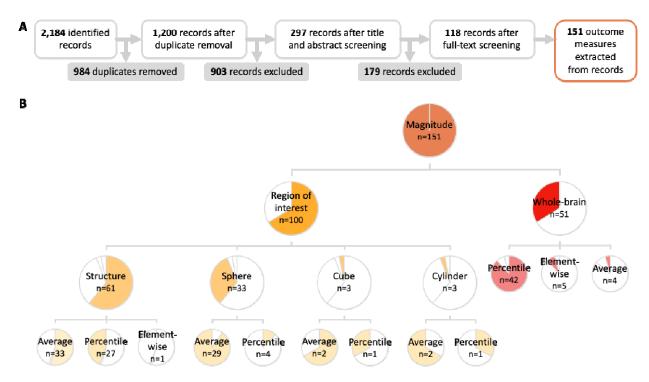


Figure 2. Extracted systematic review data. **A)** Flowchart of literature search and data screening. **B)** Tree diagram with pie charts displaying the frequency of each E-field magnitude outcome measure, relative to other outcome measures on the same level.

Table 1. Electric field magnitude outcome measures							
Group	Level 1	Level 2	tES	TMS	Reference		
Region of Interest	Cylinder	Average	1	1	[10, 20]		
		Percentile	1	-	[20]		
	Cube	Average	2	-	[19, 45]		
		Percentile	1	-	[46]		
	Sphere	Average	16	13	[4, 6, 11, 17, 18, 22, 26, 30, 44, 47-64]		
		Percentile	3	1	[11, 21, 65]		
	Structure	Average	27	6	[7, 9, 25, 66-94]		
		Element-wise	-	1	[95]		
		Percentile	20	7	[15, 43, 73-75, 89-91, 93, 96-112]		
Whole- Brain	Average		3	1	[19, 113, 114]		
	Element-wise		2	3	[8, 115-118]		
	Percentile		11	31	[4, 5, 10, 19, 20, 23-25, 30, 44, 47, 49, 61, 85, 106, 108, 113, 114, 119-140]		

 Table 1. Electric field magnitude outcome measures

3.2. Region of Interest Outcome Measures Overview

ROI outcome measures involve extracting the E-field magnitude from within a ROI. First, we discuss spherical ROIs, as insights from these ROIs yield implications for the other ROIs.

3.2.1. Spherical Region of Interest: Qualitative Results

In total, 33 studies extracted E-field magnitude via a spherical ROI (**Figure 2B** and **Table 1**). While most studies extracted the mean E-field magnitude from within the spherical ROI (n=29), a minority of studies extracted a percentile value from the spherical ROI (n=4).

A strength of spherical ROIs compared to structural ROIs and whole-brain approaches is that approximately the same area can be analyzed in terms of shape and volume across different brain regions, individuals, and studies. This facilitates a fairer head-to-head comparison of E-field magnitudes. Additionally, a spherical ROI is flexible in that its location can be derived from functional data (e.g., a hotspot obtained via TMS). However, a drawback is that spherical ROIs do not take cytoarchitectural data or complex neuroimaging data, beyond a single coordinate, into account. Also, while the spherical ROI may be identical in volume across individuals, the function and/or neurophysiology of the included brain volume can differ substantially. Likewise, defining the radius of a sphere is often arbitrary, leading to a wide variety in used radii, ranging from 0.5 to 45 mm in the included studies [21, 22]. Since each tES and TMS montage stimulates a different volume of grey matter tissue, the varying sizes of radii may or may not capture the actually stimulated network. Furthermore, as the grey matter tissue receiving the highest E-field magnitude can differ across persons within the same montage (cf., **3.3.2. Percentile-Based Whole-Brain Outcome Measures: Quantitative Results**), spherical ROIs may capture the peak E-field magnitude in different degrees across different persons.

3.2.2. Spherical Region of Interest: Quantitative Results

To assess how spherical ROI size impacts the obtained E-field magnitude, and, therefore, the interpretation of stimulation efficacy across tES and TMS protocols, we extracted the mean E-field magnitude and 10th to 99.9th percentile E-field magnitude in a percentile stepwise manner from a spherical ROI with radii ranging between 0.5 and 45 mm (**Figure 3**). Within each montage, we also assessed to what extent E-field magnitudes obtained by spherical ROIs with different sizes correlate to one another via Spearman's correlations.

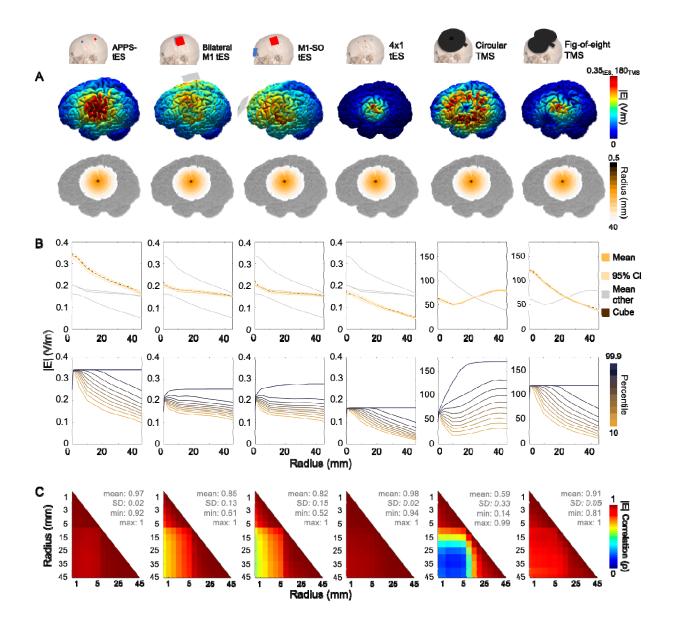


Figure 3. Effect of spherical region of interest (ROI) (radius in mm) size on E-field magnitude. **A)** Different montages with the induced E-field magnitude in an individual and the different spherical ROIs. **B**) Mean (upper row) and 10^{th} to 99.9^{th} percentile E-field magnitude (lower row) obtained by extracting the E-field in a sphere with radii ranging from 0.5 to 45 mm. In the upper row, we also show the mean E-field magnitude extracted by cubical ROIs with identical volumes to the spherical ROIs per radius size. **C**) Matrix of Spearman's correlations between obtained mean E-field magnitudes across radii with varying sizes. SD = standard deviation.

ROI size affects the obtained E-field magnitude. This is crucial for future studies aiming to disentangle whether there is an optimal E-field magnitude to attain optimal behavioral and/or neurophysiological effects, as the answer strongly depends on the ROI size. An E-field magnitude obtained via a spherical ROI with a certain radius does not equal the same E-field magnitude obtained via a spherical ROI with a different radius. Therefore, the discussion must be limited to a particular ROI location and volume when considering the dose-response relationship between E-field magnitude and clinical effects.

For tES, larger ROIs resulted in lower E-field magnitudes (**Figure 3B**). Importantly, the effect of ROI size on E-field magnitude was montage-specific. The mean E-field magnitude obtained from more focal tES montages (i.e., APPS-tES and 4x1 tES) was most affected by the size of the ROI. Conversely, the peak E-field 99.9th percentile remained stable in these montages across ROI size. For less focal tES montages, the obtained mean E-field magnitude depended less on ROI size. Here, the peak 99.9th percentile value became stable for spheres with radii exceeding 10 (bilateral M1 tES) and 20 mm (M1 – SO tES), suggesting that the peak E-field magnitude was not situated in the proximity of M1.

The differential susceptibility of more vs. less focal tES montages to ROI size can lead to interpretational pitfalls when comparing both tES classes. To illustrate, using a 5 mm radius

sphere placed at the intended stimulation target of M1, APPS-tES induces considerably larger E-field magnitudes (mean \pm SD = 0.32 \pm 0.07 V/m), compared to bilateral M1 tES (0.19 \pm 0.03 V/m). In contrast, when using a 45 mm radius ROI sphere, this difference is strongly attenuated (mean E-field magnitude APPS-tES: 0.17 \pm 0.03 V/m, bilateral M1-tES: 0.15 \pm 0.02 V/m). Thus, by varying the ROI size, one could either conclude that APPS-tES induces E-fields that are either 68% or 13% stronger than bilateral tES at the same stimulation intensity.

Reassuringly, within-montage correlation between E-field magnitudes extracted with ROIs with differing sizes was high, achieving mean Spearman's $\rho > 0.80$ for all tES montages (**Figure 3C**). The lowest correlation was found for a spherical ROI with a 0.5 vs. 45 mm radius in the M1-SO montage ($\rho = 0.52$). This implies that, for within-subject designs, the impact of different ROI sizes seems negligible when using tES in most cases.

For TMS, the effect of ROI size on E-field magnitude depends on the coil type (**Figure 3B**). The slope of E-field magnitude induced by figure-of-eight TMS over radius resembles that of focal tES montages. For circular TMS, we observed an increase in mean E-field magnitude with ROI size, when ROI size exceeded 10 mm. This, combined with the observation that peak E-field values within the ROI also increase when larger ROIs are used, emphasizes that circular TMS does not produce maximal E-field magnitudes at the coil center, resulting in small ROIs not capturing the peak E-field magnitude.

The choice of sphere size affects the comparison of both TMS modalities. Using spheres with smaller radii (i.e., ≤ 20 mm), one would conclude that figure-of-eight TMS induces higher E-field magnitudes. Using spheres with larger radii (i.e., ≥ 30 mm), one would conclude that circular TMS induces higher magnitudes. Thus, similar to tES, whether there is a necessary or optimal E-field for clinical response is contingent on not only the stimulation parameters and location, but also on the specific outcome measure used.

The within-montage correlations of E-field magnitudes obtained across different spherical ROIs were high for figure-of-eight TMS ($\rho > 0.81$, for all correlations) (**Figure 3C**), implying that the choice of ROI size will not detrimentally affect between-subject comparisons for a given montage. However, correlations were weaker for circular TMS, where a minimal correlation coefficient of $\rho = 0.14$ was found for the correlation between E-field magnitude obtained by a 2 mm vs. 35 mm sphere. This implies that, even when measured within the same subject, there is barely a relationship between E-field magnitudes obtained by ROIs with different sizes.

Overall, these findings reveal that ROI size and mean- vs. percentile-based ROI methods are critical factors to consider in the E-field modeling domain. Particularly in the search of an optimal E-field strength, ROI size should be taken into account. Smaller ROIs placed at the intended stimulation target tend to result in larger mean E-field magnitudes, but often do not encapsulate the peak E-field magnitude as well as larger ROIs, particularly in non-focal montages. Lastly, aside from circular TMS, different ROI sizes are generally well-correlated to each other, which is reassuring for between-subject comparisons.

3.2.3. Structural Region of Interest: Qualitative Results

In total, 61 studies extracted the E-field magnitude from a structure, defined as a geometrically irregular shape typically construed as an anatomical atlas or via neuroimaging data (e.g., Brodmann area or functional MRI activation map). The majority of these studies quantified E-field magnitude as the mean value (n=33) or a percentile value within the ROI (n=27).

The major advantage of structural ROIs is their flexibility towards specific research hypotheses or data. These ROIs allow to take the brain region's cytoarchitecture into account via atlases such as the Human Connectome multi-modal Parcellation atlas, or can use neuroimaging data to define ROIs in a task-, subject- or group-specific manner [141]. Thus,

structural ROIs can be tailored to specific research questions and data, whereas other ROIs may take different amounts of study-relevant vs. -irrelevant brain volumes into consideration across participants. When using structural ROIs, one should be aware that ROI size may influence the obtained E-field magnitude, in line with **Figure 3B**. However, when these ROIs are defined based on neuroimaging data, this is not necessarily a drawback.

Structural ROIs may also pose limitations beyond potentially overlooking other areas where relevant E-fields were generated. The uniqueness of structural ROIs in terms of size and shape, depending on the used data/atlas and the investigated brain region, undermines between-study and -region comparability. For instance, extracting E-field magnitudes via a structural ROI defined through a specific atlas will only provide insights with regard to that atlas and brain region. Therefore, the obtained E-field magnitude values are limited in terms of transferability to other studies, when different data or atlases are used. Moreover, as aforementioned in **2.2.3. Qualitative and Quantitative Assessments of Outcome Measures**, due to the geometric irregularity of structural ROIs, direct comparisons of different volumes, beyond what was achieved in **Figure 3**, are not meaningful due to their limited applicability.

3.2.3. Cubical And Cylindrical Region of Interest Outcome Measures

Besides the structural and spherical ROIs, a minority of studies (n = 3 for both approaches) used cubical and cylindrical ROIs. While these methods resemble the spherical ROI in that they consist of constructing a geometrically regular shape around a given coordinate, they introduce an additional sensitivity to angular placement. Still, when combined with voxel-based neuroimaging, there can be compelling reasons to use a cubical ROI. For instance, Nandi et al. (2022) used a cubical ROI to extract E-field magnitude as this ROI matched the shape and size of the voxel they used for their magnetic resonance spectroscopy analyses [19].

As the cubical ROI can be useful, we compared this ROI to its spherical counterpart. We matched the volume of each spherical ROI (e.g., to match the 1 cm radius sphere, which has a volume of 4.189 cm³, we established a cube with a length, width and height of 1.612 cm) and extracted the mean E-field magnitude from the grey matter volume within our cubical ROIs, centered along the X, Y and Z axes. Our results (**Figure 3B**) indicate that when matched for volume, cubical and spherical ROIs result in close-to-identical mean E-field magnitudes. Thus, the same principles that we discussed for spherical ROIs should also apply to cubical ROIs.

3.3. Whole-Brain Outcome Measures Overview

In contrast to the regional specificity inherent to ROI analyses, whole-brain approaches remain agnostic to the spatial location receiving stimulation. In doing so, these analyses mitigate the risk of overlooking unexpected brain regions that receive relevant amounts of stimulation. This is a major strength assuming that the effects of tES and/or TMS are governed by a (monotonic) dose-response relationship in which a key factor is the peak E-field magnitude in any brain region.

3.3.1. Percentile-Based Whole-Brain Outcome Measures: Qualitative Results

Most studies (n=42) used a percentile approach to quantify whole-brain E-field magnitude. This mitigates the risk of spatially poorly defining a ROI and overlooking the location that received the maximal E-field magnitude, potentially increasing the consistency of measured effects across participants and studies (**Figure 3B**). On the other hand, only extracting a percentile from E-field simulations introduces regional uncertainty. That is, the percentile approach is spatially undefined, resulting in it potentially providing information about unexpected regions that are not directly the stimulation target or providing information about different regions across participants.

3.3.2. Percentile-Based Whole-Brain Outcome Measures: Quantitative Results

To disentangle how different percentile cut-offs affect the obtained E-field magnitude, we extracted the 10th to 99.9th percentile in a 10% incremental stepwise manner for all participants and montages. We also extracted and visualized the volume analyzed per percentile.

Although all approaches analyzed nearly identical volumes of grey matter tissue across the different percentiles (**Figure 4B**), visualization of the analyzed regions highlights that substantially different brain regions were analyzed by high percentile cut-offs across different tES and TMS modalities. For instance, **Figure 4A** indicates that the region receiving the highest E-field magnitude as a result of non-focal tES modalities is not located in M1. Perhaps even more concerning is the difference in analyzed regions across participants. **Figure 5** shows that the regions analyzed particularly by the 99.9th percentile, can differ tremendously across persons, particularly in the non-focal tES montages. For instance, participant 88 (S88) shows peak E-fields in the proximity of M1 as a result of bilateral tES but not M1-SO tES, whereas participant 99 (S99) shows the opposite pattern. Thus, even the same outcome measure and threshold value can report E-field magnitudes in substantially different regions in a person-specific manner.

For both TMS and tES, there was a steep increase in E-field magnitude from the 90th to 99.9th percentile (**Figure 4B**). E-field magnitude more than doubled for focal modalities, and increased by over 50% for non-focal modalities.

Overall for tES, the E-field magnitudes obtained by percentile approaches correlated strongly, with the mean correlation exceeding $\rho = 0.86$ for all modalities (**Figure 4C**). This is reassuring, as it implies that using different percentiles will likely not vastly impact conclusions of studies comparing between subjects. The lowest correlation value included the 99.9th percentile E-field magnitude across all tES modalities, with 4x1 tES showing the lowest correlation ($\rho = 0.46$) between the 99.9th and 10th percentile. For TMS, the correlation between the 99.9th percentile vs. all other percentiles is considerably lower than the correlation between the tES 99.9th percentile vs. all other tES percentiles. Likely, this discrepancy stems from the

fact that the magnetic field strength is more prone to decay with distance, while tES induced fields are governed by electro-conductive principles (**Figure 4B**).

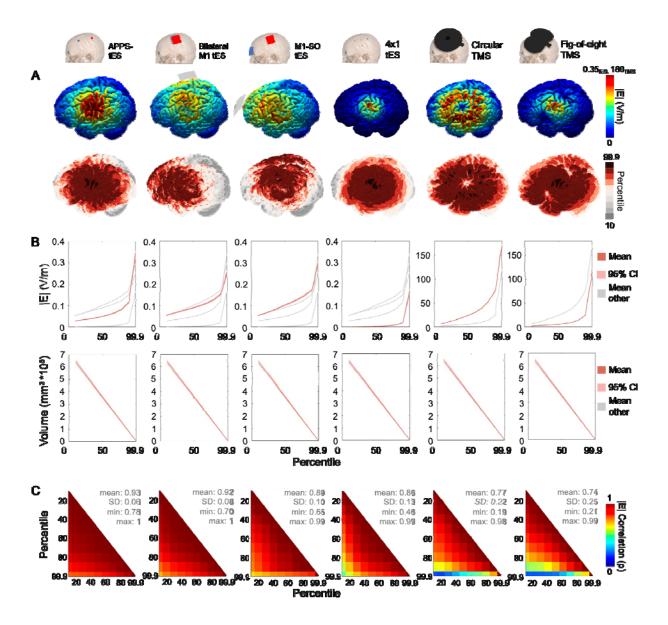


Figure 4. Effect of different percentiles on extracted E-field magnitude on the whole-brain level. **A)** Different montages with the induced E-field magnitude in an individual and the regions included in the different percentiles. **B**) 10th to 99.9th percentile E-field magnitude (upper row) and included volume (lower row). **C**) Matrix of Spearman's correlations between obtained E-field magnitudes across varying percentiles.

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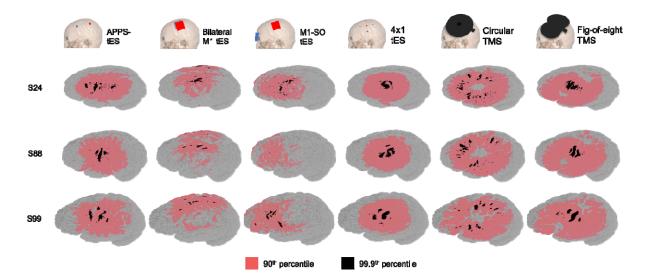


Figure 5. Volumes analyzed by the percentile-based outcome measure across all modalities in three subjects (S24, S88, and S99). Although the volumes analyzed by the 99.9th percentile differ considerably for all modalities, this difference is most remarkable for bilateral and M1-SO tES.

3.3.3. Element-Wise Whole Brain Outcome Measure

A minority of studies using a whole-brain approach extracted the element-wise E-field magnitude (n=5). While this method resembles percentile-based extraction, it is biased towards the spatial resolution of the head model (i.e., the number of tetrahedra in the case of SimNIBS). To illustrate, assuming that each grey matter element is identical in volume and the distribution of E-field magnitudes across elements is uniform, extracting the E-field magnitude of the 100th highest element will result in a value similar to the 50th percentile in a model with 200 elements (low spatial resolution) and a value similar to the 90th percentile in a model with 1,000 elements (high spatial resolution).

3.3.4. Mean Whole Brain Outcome Measure

Four studies extracted the mean E-field magnitude in the whole grey matter. Using this approach yields the disadvantage that, certainly for focal types of stimulation, it does not represent the E-field magnitude to which some regions were exposed.

3.4. Direct Comparison of Different Electric Field Magnitude Outcome Measures

To meaningfully inform future work, we directly compared two of the most common outcome measures: the mean spherical ROI approach and percentile-based whole-brain approach. To do so, we examined the difference and correlation between the obtained E-field magnitudes and the overlap between the analyzed brain volumes. For the spherical ROI, we used radii ranging between 0.5 and 45 mm. For the percentile-based whole-brain analyses, we used percentiles ranging between the 10th and 99.9th percentile.

First, we assessed to what extent the ROI and whole-brain approaches retain similar Efield magnitudes via Spearman's correlations and E-field magnitude differences (%). While correlations inform how E-field magnitudes relate to the effects of stimulation across participants, E-field magnitude difference is a key consideration for research investigating absolute E-field magnitudes. Overall, correlations were highest for APPS-tES, where the mean correlation value was $\rho = 0.86$ (range = 0.76–0.98) and the lowest correlation value of $\rho = 0.76$ was obtained for the correlation between the 10th percentile and a 4 mm spherical ROI. For bilateral M1-tES, the mean correlation value was $\rho = 0.72$ (range = 0.35–0.97). The highest correlations were present between the 90th percentile and the larger spheres (i.e., spheres with radii ≥ 30 mm). With these comparisons, all correlations exceeded $\rho = 0.95$. For M1-SO tES, the mean Spearman's value was 0.56 (range = 0.15–0.82); correlations were highest between the 80th and 90th percentile and spherical ROIs with radii exceeding 15 mm. Here, all correlation values exceeded $\rho = 0.70$. Notably, for M1-SO tES, the correlation between the 99.9th percentile E-field magnitude and the spherical ROI was low-to-moderate, irrespective of ROI size (highest correlation: $\rho = 0.54$, lowest correlation: $\rho = 0.15$). Regarding 4x1 tES, the mean correlation value was $\rho = 0.77$ (0.42–0.99). The obtained correlation values resembled those of APPS-tES, although they were generally lower. Regarding circular TMS, the mean correlation value was $\rho = 0.23$ (range = -0.13–0.84). The highest correlations for circular TMS were present between spherical ROIs with radii ≥ 40 mm (lowest correlation: $\rho = 0.83$) and the 99th percentile. Overall, circular TMS showed the weakest correlations between spherical E-field magnitudes and percentile-based E-field magnitudes, further corroborating that this type of TMS does not induce maximal E-field in the targeted ROI. For figure-of-eight TMS, the mean correlation value was $\rho = 0.33$ (range = -0.04–0.96). Here, a similar trend to the other more focal montages was present (i.e., high correlation values for the 99.9th percentile against spherical ROIs of all sizes, and a gradual decrease in correlation values along with decreasing percentiles), although the obtained correlation values were notably lower for TMS vs. tES, pointing to a key consideration in future study designs depending on stimulation type.

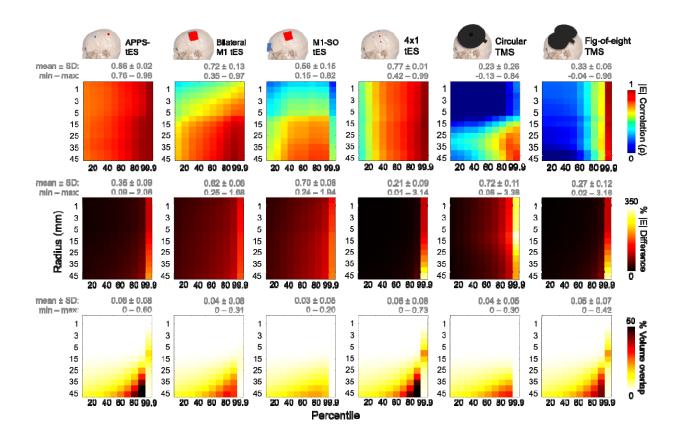


Figure 6. Comparison of spherical ROI and percentile-based whole brain E-field magnitude approaches in terms of correlation (upper row) and difference (middle row) in obtained E-field magnitudes and overlap in analyzed volume (bottom row).

Concerning E-field magnitude difference, the discrepancy was largest between E-field magnitudes in the 99.9th percentile against all ROIs. Here, differences of up-to 338% were present (circular TMS, 99.9th percentile vs. 10 mm ROI sphere). Although the differences with ROI-obtained E-field magnitudes were smaller for the other percentiles, the large overall difference between both approaches emphasizes the importance to only interpret and compare E-field magnitudes across studies using similar outcome measures.

The overlap of analyzed grey matter volumes is shown in **Figure 6**, **bottom row**. Overlap was generally low, with the mean overlap across all percentiles and ROI spheres not exceeding 6%. The highest overlap was present between the 90th percentile and the large spherical ROIs with radii exceeding 30 mm. Here, an overlap between analyzed volumes of up to 60% (APPS-tES), 73% (4x1 tES) and 42% (Figure-of-eight TMS) was present. Of note, using ROIs of 10 mm yielded the greatest volumetric overlap with the 99.9th percentiles, obtaining values of \leq 21% overlap. **Figure 7**, which visualizes three spherical ROI volumes and two percentile whole-brain volumes further adds to this interpretation. Here, we show that the volume analyzed by the 90th percentile indeed best resembles that of larger spherical ROIs. Moreover, based on the 99.9th percentile, it becomes clear that the less focal types of stimulation induce their peak E-field magnitude outside of the intended stimulation target, while the spatially defined spherical ROI analyzes the E-field at the user-defined target.

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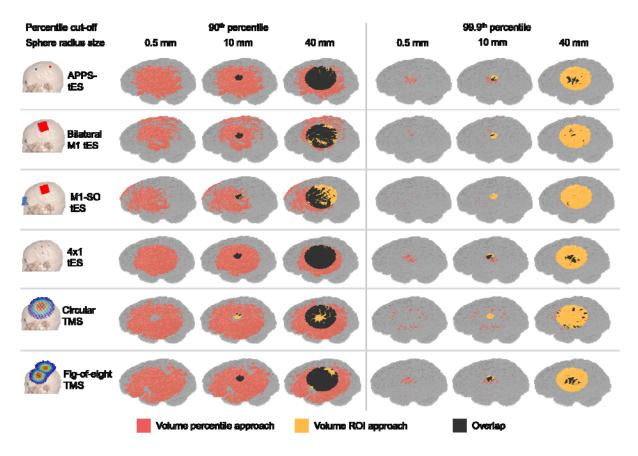


Figure 7. The volume analyzed by the spherical ROI approach (yellow), the percentile approach (red), and both approaches (i.e., the overlap) (black) in a representative participant.

4. Discussion

In this combined systematic review and E-field modeling study, we investigated how the selected outcome measure impacts E-field magnitude. In the 151 experiments reporting E-field modeling results that fit our systematic review criteria, outcome measures fell into two major categories: ROI (n = 100) and whole-brain approaches (n = 51). While most tES studies used the ROI approach (71% of studies), the whole-brain approach was used by the majority of TMS studies (68.6% of studies). We computed 600 E-field models in 100 healthy younger adults to further examine how E-field magnitude outcomes alter the interpretation of tES and TMS models that are more or less focal in stimulation diffusivity. Collectively, we computed 106,800 E-field magnitude related outcome measurements, finding that the selected outcome measure

significantly impacts the interpretation of within- and between-subjects modeling. Specifically, for ROI-based outcomes, the wide range of sphere radii used in the literature examines volumes with a resolution level spanning from gyri to close-to-whole hemispheres. ROI size also interacts with stimulation focality; whereas the peak stimulation effects of focal forms of tES (APPS-tES and 4x1 tES) and TMS (figure-of-eight coil) are captured with focal ROIs placed at the stimulation target, the peak effects of less focal forms of tES (bilateral M1 and M1-SO) and TMS (circular coil) are better encapsulated with larger ROIs since these modalities do not always position the peak E-field at the intended target.

Regarding whole-brain approaches, we compared the E-field magnitude for each stimulation modality using percentiles ranging from the 10^{th} to the 99.9^{th} percentile in 10-percentile increments, finding high Spearman's correlation values between each percentile with a mean value of at least $\rho = 0.74$ for all modalities. However, we found that percentile-based approaches analyzed different brain regions depending on the participant and the montage, introducing spatial uncertainty as a potential pitfall when interpreting the obtained E-field magnitudes. Namely, in less focal montages such as M1-SO tES, extracting E-field magnitudes through a high percentile threshold can result analyzing entirely different brain regions across persons. Therefore, it is important to consider how E-field outcome measure selection can not only alter group-level data but also data on the individual level.

The key question of whether there is an optimal E-field magnitude for dose-response relationships must be discussed within the context of each specific measure as we demonstrated wide variance of E-field modeling results based on the selected E-field measure. For instance, the optimal E-field for the 99.9th percentile almost certainly differs from that at the 10th percentile due to the volume examined and the consideration of different brain regions depending on the ROI size. Thus, more consistency and attention to E-field magnitude reporting measures is needed across studies.

We computed a percentile volumetric overlap between all 140 possible combinations per participant and modality to directly compare the mean ROI and percentile whole-brain approach. Strikingly, the mean overlap of percentile and ROI based outcome measures did not exceed 6%, suggesting that whole-brain percentile and spherical ROIs report the E-field magnitudes from fundamentally different brain regions. Overall, the low overlap underscores the critical importance of researchers selecting suitable E-field outcome measures to appropriately investigate their research questions. As we demonstrated here, selecting a suitable E-field outcome measure depends on the focality of the stimulation approach, the targeted region, and the volume of the targeted region or network. Crucially, these data directly comment on why it is important to be mindful when comparing across E-field modeling studies; even within the same person, stimulation parameters, and target, vastly different E-field magnitude findings can be extracted with mean ROI vs. percentile ROI vs. whole-brain percentile approaches. A valid critique is that the 6% overlap between the ROI and whole-brain percentile outcomes collapses across different volumes and therefore this overlap may be expected to be low. However, even when considering similar volumes between the ROI and whole-brain percentile approaches (e.g., the 90th percentile and large spherical ROIs with radii exceeding 30 mm), the highest overlaps between outcome measures were still relatively low (60% for APPS-tES, 73% for 4x1 tES, and 42% for figure-of-eight TMS). Thus, even in the best case scenario, when comparing E-field magnitude results from ROI and whole-brain percentile approaches, analyzed volumes from at least 27% of the regions did not overlap.

4.1. Recommendations For Future Research

It is clear that the selected outcome measures can substantially affect the obtained Efield magnitude, and therefore, the interpretation of the results of computational simulations. The heterogeneity between outcome measures differs not only between ROI and whole brain approaches but also within measures such as the radius of a spherical ROI varying between 0.5–45 mm or the percentile cutoff varying between the 10th to 99.9th percentile. It is likely that these results are also translatable to the subcomponents of E-field magnitude (i.e., normal and tangential E-field strength), which use similar outcome measures.

Although ultimately, an outcome measure is always chosen in the context of specific research goals, there are some recommendations that can be made based on our review and data analyses that should be taken into account when deciding which outcome measure to use.

Recommendation #1: Use the Study Goal to Select the Most Suitable Outcome Measure

First, the choice for an outcome measure depends on the goal of the study and the available data. If neuroimaging data such as functional MRI is available, defining a structural ROI based on this data may be a viable approach. If the study's goal is to compare E-field magnitudes in different neural regions or across persons, a spherical ROI may be best suited due to it analyzing nearly-identical grey matter volumes. It is important to weigh the pros and cons of the ROI and whole-brain percentile approaches as briefly summarized below and in **Table 2**.

One characteristic that is inherent to all ROI outcome measures is the spatial confinement of these approaches. This is both a strength and weakness. It can be compelling when one is interested in a specific brain region. Conversely, this spatial confinement yields the danger that it is indifferent to the actual location that received maximal stimulation, which may not always coincide with the selected ROI but can play an important role to explain behavioral and/or neurophysiological effects. In addition, structural ROIs are highly unique which may impede between-region, -subject and/or -study comparisons and when using atlas-based structural ROIs, one may not be able to isolate the subregion of interest. On the other hand, spherical ROIs are biased towards the size of the sphere, particularly when the mean E-field magnitude is extracted, and may incorporate different volumes of task relevant grey matter across individuals.

Whole-brain percentile approaches may be best suited when researchers are interested in knowing the peak E-field magnitude irrespective of whether this coincides with the intended stimulation target. An advantage of whole-brain percentile outcome measures is that they produce identical outcomes by virtue of the mathematical threshold used, whereas some prior studies have used user-defined ROIs that could be prone to low inter-rater reliability. However, a limitation of the percentile approach is that it does not confirm target engagement of the intended stimulation region. Rather, with more diffuse stimulation methods such as M1-SO tES, the peak E-field is induced midway between electrodes. Thus, spherical ROI and whole-brain percentile approaches each have pros and cons and the correct choice depends on the researcher's question.

Table 2. Pros and cons of the most common electric field modeling outcome measures							
	Structural ROI	Spherical ROI	Whole-brain percentile				
	Confined: Defining a I magnitudes within a p	ROI facilitates interpretation of E-field predefined brain area.	<u>Unconfined</u> : By considering the whole-brain, information is always given about the peak E-field magnitude regardless of location				
Pros	<u>Highly flexible</u> : The ROI can be tailored to available	Flexible: The center of the ROI can position based neuroimaging data and/or research questions.					
	neuroimaging data and/or specific research questions	Transferrable: Similar volumes across brain regions, participants or montages are analyzed.	<u>Reproducible</u> : Easy to replicate as only a single percentile value is required to obtain the same results				
Cons	<u>Confined</u> : By focusing fields outside of the R	Spatially uncertain: Different					
	<u>Transferrable</u> : Uniqueness may hinder transferability	Size: Defining ROI size can be arbitrary, but can strongly affect the obtained E-field magnitude.	brain volumes and regions may be analyzed across montages and participants. This can also impede interpretation of the obtained E-field magnitude.				
	to other brain regions, participants and/or studies	Included region: Different cytoarchitectural / functional regions may be included in the same ROI across persons					

Table 2. Pros and cons of the most common electric field modeling outcome measures

E-Field = electric field, ROI = Region of Interest

Recommendation #2: Consider the Noninvasive Brain Stimulation Modality and Focality

Second, the optimal choice of outcome measure depends on which noninvasive brain stimulation modality is used. While ROI-based outcome measures are an excellent way to quantify the E-field magnitude in a specific region, their value is highest when analyzing more focal types of stimulation. In contrast, when less focal stimulation types are used, spatially confined ROIs placed directly underneath the electrodes or the coil center will often exclude the actual region that received the highest E-field magnitudes as the region of peak E-field intensity often differs from the target region and may differ across individuals.

<u>Recommendation #3: The Dose-Response Relationship Between E-Field Magnitude and</u> Clinical Outcome Must Only Be Compared Within a Singular Outcome Measure

A primary finding of our systematic review is that researchers have used many different approaches to quantify E-field magnitude. With the powerful tool of E-field modeling, a question that many of us seek to answer is whether there is a positive dose-response relationship between the induced E-field magnitude and clinical response. It is enticing to consider these results monolithically and pursue an elegant statement such as, "The optimal E-field is V/m." The simplicity of pursuing a singular value is appealing and would enable easier dissemination of individualized E-field dosing such as through dose-controlled tES, 2-Sample Prospective E-field Dosing (2-SPED), or applying an individualized TMS dose to induce a singular E-field at the cortical level across individuals [49, 51]. However, there are many reasons to believe that a singular optimal E-field value does not exist and that more nuance is necessary. For instance, a singular optimal E-field value would likely not apply in the same way across different brain regions due to varying neuronal populations, white matter tracts, and grey matter densities, among other variables [141-144]. Age, sex, diagnosis, and other typical demographical considerations likely also impact the optimal E-field dose [39, 64, 133, 145, 146]. Furthermore, the dynamic nature of the brain might further complicate things, as time-varying changes within a brain region of a single person may further modify the optimal E-field dose

[147, 148]. Here, we added outcome measure as a key consideration to take into account in any discussion of optimal E-field dose. The E-field magnitudes extracted from the same models widely vary depending on the volume and regions considered. Thus, instead of the monolithic goal of a singular optimal E-field, we might instead work towards more nuanced goals taking many factors into consideration. In the future, we might come to the understanding that, "The optimal mean E-field for TMS, in a 5 mm radius spherical ROI centered over the motor hotspot as defined by TMS, in 50- to 70-year-old adult patients with ataxia, is _____ V/m." This value would almost certainly be different than an "optimal prefrontal E-field magnitude for TMS, measured by the 95th percentile whole-brain approach, and in 20- to 40-year-old adult patients with depression." Of course, additional refinement might further personalize our understanding of any "optimal" E-field value between individuals even when they have similar ages or diagnoses.

Recommendation #4: Time for Standardized E-Field Outcome Measure Reporting

Pursuing goals such as better understanding the dose-response relationship between Efield magnitude and therapeutic outcome necessitates that researchers report more standardized outcome measures between studies. As highlighted in <u>Recommendation #3</u> and throughout this study, comparing most ROI and whole-brain percentile measures is akin to comparing apples to oranges. Strikingly, it is even dissimilar to compare within some of these outcome measures across modalities, and even within the same participant and same E-field model when the ROI size has varied or the percentile has changed. Thus, in order to work toward more suitable comparisons across studies, it is important to consider how to improve the consistency of reporting across studies.

Notably, this recommendation does not conflict with <u>Recommendation #1</u> calling for researchers to choose the E-field outcome measure that best suits the experimental question.

Rather, these reporting standards can apply to specific experimental questions which require different modeling outcome measures.

While deriving a comprehensive list of standard reporting procedures is beyond of the scope of this study and warrants a consensus-based approach, we propose a short list of preliminary guidelines to adopt across studies:

- To state that a specific brain region was stimulated, researchers must use a ROI-based method and describe how the ROI was defined in the Methods section. We propose this guideline since the peak E-field intensity derived from the whole-brain percentile approach does not always coincide with the intended stimulation target, particularly with less focal forms of brain stimulation.
- 2. When defining a ROI, we recommend that researchers report the MNI coordinate that the ROI is centered on. In cases such as structural ROIs in which the researcher might individually define the ROI, an average MNI value should be provided when possible to aid in the reproducibility of findings and comparisons between studies.
- 3. Whether using a ROI or whole-brain percentile approach, the volume of the examined tissue should be reported. This recommendation seeks to allow for some degree of comparison between ROI and whole-brain percentile approaches since the reader should at least be able to determine whether a similar volume was analyzed. Researchers should ideally also visualize which regions are analyzed by a certain approach, with this being particularly important for percentile-based approaches to enable the reader to interpret the regions of the extracted volume.
- 4. Finally, we recommend all future studies to measure and report multiple outcome measures whenever possible. For instance, research applying M1-SO tES could use both the spherical ROI approach if one is interested in a specific region, and the percentile-based approach with complementary visualizations, to assess whether peak E-fields were induced in the intended region.

Table 3. Overview of E-field modeling outcome measure related recommendations

Recommendation

- #1 Use the study goal to select the most suitable outcome measure
- #2 Consider the noninvasive brain stimulation modality and focality
- #3 The dose-response relationship between e-field magnitude and clinical outcome must only be compared within a singular outcome measure
- #4 Time for standardized E-field outcome measure reporting
 - #4.1. Researchers must use a ROI-based method to state that a brain region was stimulated
 - #4.2. When defining a ROI, researchers should report a representative MNI coordinate
 - #4.3. The volume of the examined brain area should be reported
 - #4.4. Studies report multiple outcome measures whenever possible.

4.2 Conclusions

Outcome measures in the computational noninvasive brain modeling field have received minor attention in the past. Based on our systematic review and computational modeling of 106,800 E-field magnitude outcome measures across 100 participants and 600 tES and TMS E-field models, we show that different outcome measures substantially affect the obtained E-field magnitude and the analyzed brain region in a montage and person-specific manner. Therefore, one should only interpret and compare E-field magnitudes across studies when similar outcome measures are used. We formulated four recommendations (**Table 3**) for future research to ensure the informed selection of outcome measures. Our hope is that adopting these recommendations will enable future studies to avoid interpretational pitfalls and reduce the inconsistency of the used E-field outcome measures.

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Conflicts of interest

We confirm that all authors have no known conflicts of interest or competing interests associated with this publication and there has been no financial or personal relationship with other people / organizations that could inappropriately influence this work.

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