Wanding Through Space: Interactive Calibration for Electric Muscle Stimulation
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Wanding Through Space: Interactive Calibration for Electric Muscle Stimulation

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ABSTRACT

Electric Muscle Stimulation (EMS) has emerged as an interaction paradigm for HCI. It has been used to confer object affordance, provide walking directions, and assist with sketching. However, the electrical signals used for EMS are multi-dimensional and require expert calibration before use. To date, this calibration has occurred as a collaboration between the experimenter, or interaction designer, and the user/participant. However, this is time-consuming, results in sampling only a limited space of possible signal configurations, and removes control from the participant. We present a calibration and signal exploration technique that both enables the user to control their own stimulation and thus comfort, and supports exploration of the continuous space of stimulation signals.

CCS CONCEPTS
• Human-centered computing → Haptic devices; Gestural input; • Hardware → Emerging interfaces;

KEYWORDS

Electric muscle stimulation, EMS, functional electrical stimulation, calibration, haptic feedback

ACM Reference Format:

1 INTRODUCTION

Electric muscle stimulation (EMS) and Functional electric stimulation (FES) have seen increased use in HCI. Both use electrical signals to stimulate a user’s nerves and muscles, either to provide haptic feedback or to cause movement. However, this stimulation requires careful calibration of many signal parameters, such as frequency, pulse width, or amplitude, as these influence what kind of sensation is evoked in the user. Furthermore, individual differences in skin thickness or subcutaneous fat necessitate per-user calibration.

Existing calibration techniques, especially in HCI, either use pre-selected pulse widths and frequencies, and thus only calibrate for amplitude, or explore only a small number of discrete pulse-widths and frequencies, alongside the amplitude. Whichever of these techniques is adopted, they both result in an exploration of only a limited part of the signal parameter space. Furthermore, these calibration techniques are time-consuming and rely on a complex collaboration between the experimenter and the participant. Typically, the experimenter selects parameters, controls the amplitude, and waits for a reaction from the user (either physical or verbal). This process removes control from the participant, even though FES can result in uncomfortable stimulations.

We present a new calibration and exploration technique, which addresses these problems. Selection of pulse width, frequency, and amplitude parameters here is mapped to a 3D space in front of the user. The user can then explore the stimulation parameters by moving through this space. This both puts the control back in the hands of the participant, and results in a calibration based on an exploration of a much wider space of parameters, without any incurred time penalty. Through a user study, we demonstrate the calibration of movement-causing EMS and FES for haptic feedback for smartwatches.
2 RELATED WORK

FES/EMS has been used to a variety of ends—both to cause movement (e.g., [7, 12, 15, 16]) and provide haptic feedback (e.g., [5, 10]). At Augmented Human, for example, EMS has been used to change the flavor of soup by stimulating the tongue [1], add vibrato to speech [4], and to teach drumming rhythms [2].

While it is a critical part of any EMS system, calibration has received relatively little attention in HCI so far. We know from Lopes et al. [12], that calibration can be time consuming (2–5 minutes per participant per pose), and Pfeiffer et al. [15] describe the need for high levels of accuracy (electrode placement within 5mm of the optimal location). However, no standardized or generalizable approach has yet been adopted.

EMS calibration can be split into two components parts [9]: (1) spatial calibration—selecting the location for the electrodes—, and (2) signal calibration—selecting signal parameters to use (frequency, pulse width, and amplitude).

In this paper, we focus on signal calibration. This is typically approached in one of two ways: either (1) the authors pre-select the frequency and pulse width (e.g., [2, 11, 18]), and explore the amplitude with the participants. Alternatively, (2) the authors and participants explore a limited sub-range of the available parameter space (e.g., [13], where they fix the frequency, and explore pulse width and amplitude). Whichever technique is selected, only a small part of the available parameter space ends up being explored.

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As shown in Figure 2, participants explored a large portion of the available space. The heatmap shows a projection of which parts of the space were visited with crosses marking chosen notification settings. Highlighted in red are the chosen notifications from one participant. Participants detailed what they would use the notification for, and provided Likert responses about comfort and disruption. At the end of the second task, participants were asked (1) how easy it was to distinguish sensations, (2) how long it took to find their chosen sensations, and (3) their movement and selection strategies for the task.

Finally, the electrodes were moved to cover the flexor carpi radialis and palmaris longus muscles, on the inside of the forearm. The stimulation cuboid was configured with the parameters: 1–4 ms pulse width, 35–80 Hz frequency, 0–100% intensity. The participants were asked to find movement causing (muscle activation) parameters within the space, and rate them as comfortable or not-comfortable. In this way, the users specified their preferred movement causing parameters across a continuous parameter space. The experimenter recorded the participants’ ratings. At the end of this task, participants completed a questionnaire about anxiety, intuitiveness and simplicity, and were asked about strategies of movement for this task.

The study took 25 minutes on average. Participants were compensated the equivalent of $15 for participating.

## 5 RESULTS

For the analysis, we removed data from three participants (P1, P7, and P16), because they only completed parts of the study. P1 and P7 were due to experimenter error with the study interface, while P16 opted to not complete the study. We describe the results, based on the data from the remaining 14 participants, below.

### 5.1 Exploration

As shown in Figure 2, participants explored a large portion of the available space. Participants explored on average 31% of the space over 3.7 minutes. They tended to stay towards the inside of the space, as shown by the drop-off towards the edges of the graph.

Participants described exploring the stimulation space as intuitive and easy (Figure 3). The exploration did not make them anxious or nervous, and they were able to develop a good understanding of the stimulation space.

### 5.2 Notification

Figure 4 shows the participants explored 27% of the available space while picking notification parameters (on average). They placed 3–5 (M=4.4) notifications, making use of the available space (see Figure 4). Participants spent on average 6.6 minutes finding notifications. Interestingly, participants explored the space for 45 seconds on average before selecting their first notification.

Participants also rated the comfort and disruption of their selections. Higher comfort ratings were seen as less disruptive: Spearman’s rank correlation coefficient $r_s = -0.74, p < 0.0001$.

Participants defined notifications to cover a range of scenarios, including email alerts (n=5), phone calls (n=11), navigation updates (n=2), and medical warnings (such as diabetes alarms, n=3). Across these notifications, participants felt that, in general, it took them a long time to decide on sensations (Figure 5). There was no consensus about whether sensations were easy to distinguish.
Across the three study tasks, participants explored a combined 53% of the stimulation space on average. As this is a continuous space across a wide range of pulse widths, frequencies, and intensities, this corresponds to a much wider range of parameter exploration than achievable using traditional signal calibration methods in HCI. In combination with the diversity of chosen stimulation parameters (either for notifications or activation), this shows the benefits of enabling exploration across a wide space. Exploring the parameters in 3D also enables faster signal calibration. Participants specified 14.7 muscle stimulation locations in 4.37 minutes, and 4.4 notification locations in 6.6 minutes. This represents a large improvement over the 2–5 minutes per pose (where pose is equivalent to selecting one preferred location) typical with traditional techniques [12].

Exploring 53% of the stimulation space represents a large portion of the space where stimulation can be perceived, and is not uncomfortable. Anecdotally, participants were quick to identify ‘unpleasant’ locations in the 3D space, which they would then not explore further.

In this study, we have explored a 6 degree of freedom (DOF) exploration of a parameter volume. This enables participants to move the wand to a specific area to experience given stimulation parameters. However, additional degrees of freedom can be added. For example, once a parameter location has been selected, users could also specify the dynamics of the desired stimulation (i.e., the ‘ramp’ with which the stimulation is applied, from 0% amplitude up to the desired x% amplitude, over time), by rotating the wand. Rotating left and right could specify the rate at which the intensity increases. This could be similarly achieved by allowing users to specify paths through the volume, rather than specific locations. Further, the wand could be replaced by a device that combines other input modalities (such as a VR controller), this could also enable dynamic rating or clutching interactions for varying the parameters within the space.

Importantly, our calibration technique gives control to the user. This is (ethically) important when they are exploring a space of interaction that could potentially be uncomfortable. The participants did not report being anxious or nervous during the study, however, anecdotally, some participants were uneasy when being introduced to the concept of the study. We believe that being in control is a valuable mechanism to help to reduce this apprehension surrounding EMS applications.

7 CONCLUSION
Calibration is a fundamental part of any FES/EMS-based system. To date, practitioners explore only a small range or subset of parameters when performing signal calibration. Typically this involves pre-selecting a small number of parameters and applying them to a participant to gauge their feedback. This results in control being removed from the participant, all while they can be exposed to uncomfortable, or sub-optimal, stimulation. We present a 3D exploration-based calibration technique that maps a large space of stimulation parameters to a 3D volume. By moving through the 3D volume, the participants control the parameters of their stimulation and can select those comfortable for them. In a study, we demonstrate participants exploring a wider range of stimulation parameters than previously attempted in HCI, all the while maintaining full control of the resultant sensations.

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