Brief Communications

Stochastic Facilitation of Artificial Tactile Sensation in **Primates**

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Artificial sensation via electrical or optical stimulation of brain sensory areas offers a promising treatment for sensory deficits. For a brain-machine-brain interface, such artificial sensation conveys feedback signals from a sensorized prosthetic limb. The ways neural tissue can be stimulated to evoke artificial sensation and the parameter space of such stimulation, however, remain largely unexplored. Here we investigated whether stochastic facilitation (SF) could enhance an artificial tactile sensation produced by intracortical microstimulation (ICMS). Two rhesus monkeys learned to use a virtual hand, which they moved with a joystick, to explore virtual objects on a computer screen. They sought an object associated with a particular artificial texture (AT) signaled by a periodic ICMS pattern delivered to the primary somatosensory cortex (S1) through a pair of implanted electrodes. During each behavioral trial, aperiodic ICMS (i.e., noise) of randomly chosen amplitude was delivered to S1 through another electrode pair implanted 1 mm away from the site of AT delivery. Whereas high-amplitude noise worsened AT detection, moderate noise clearly improved the detection of weak signals, significantly raising the proportion of correct trials. These findings suggest that SF could be used to enhance prosthetic sensation.

Introduction

Perhaps somewhat counterintuitively, combining informative signals with noise can enhance signal processing in biological systems (Moss et al., 2004). This phenomenon is referred to as stochastic facilitation (SF) or stochastic resonance (McDonnell and Ward, 2011). SF occurs when moderate levels of noise increase the number of neural threshold crossings, but also keep those crossings in synchrony with the signal (Moss et al., 2004). More intense noise masks the signal and worsens its detection. Examples of SF in human sensory processing include improved detection of visual stimuli (Kitajo et al., 2003), as well as mechanical stimuli applied to the fingers (Collins et al., 1996) and the foot sole (Wells et al., 2005). Similarly, SF has been reported for the cercal sensory system of crickets (Levin and Miller, 1996), CA1 neurons in hippocampal slices from rats (Stacey and Durand, 2001), and the spinal circuits of the cat (Martínez et al., 2007).

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However, there is an ongoing debate as to whether SF serves any positive function in the neural circuitry (McDonnell and Abbott, 2009; McDonnell and Ward, 2011). Additionally, it is unclear whether SF could be used in assistive devices for people with sensory disabilities, for example, systems for sensory substitution (Bach-y-Rita and Kercel, 2003) and neural prosthetics with artificial sensation (Romo et al., 1998; O'Doherty et al., 2011).

Recently we described a sensorized neural prosthetic, called BMBI (brain-machine-brain interface), where neuronal activity recorded in the primary motor cortex (M1) was used to control the movements of a virtual arm while artificial tactile feedback from such an actuator was generated by intracortical microstimulation (ICMS) patterns delivered directly to somatosensory cortex (S1) (O'Doherty et al., 2009, 2011, 2012). This paradigm opens the possibility of aiding future human users of neuroprosthetic limbs with sensory signals arising from a variety of limb sensors. However, to develop sensorized neural prostheses further, detailed exploration is needed into ICMS as the means to reproduce the exteroception and proprioception normally provided by somatic sensation. Of particular interest is an inquiry into the processing of sensations evoked by ICMS when they are mixed with noisy inputs. For natural sensation, this problem is routinely solved by biological organisms in real-life environments (Moss et al., 2004). For artificial sensation, however, the research is only starting (O'Doherty et al., 2012). In the present study, we hypothesized that SF could enhance detection of weak ICMS patterns under an active exploration paradigm (O'Doherty et al., 2011; 2012). Accordingly, we delivered mixtures of stimulus and noise to S1 in different proportions and tested how their different combinations affected the subject's active exploration of artificial textures (ATs).

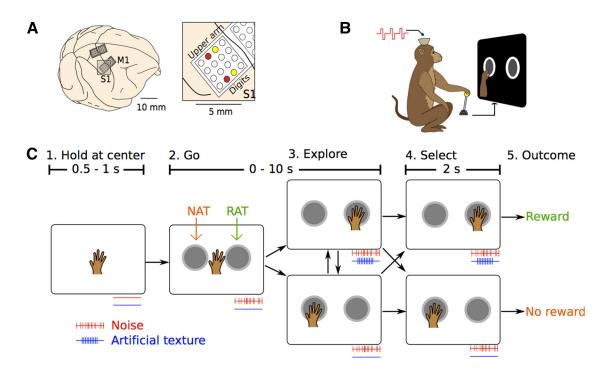


Figure 1. Experimental setup and task schematics. **A**, Arrays of microelectrodes were implanted in S1 and M1. Stimulating electrodes for monkey M (inset) are highlighted in red (AT) and yellow (noise). **B**, Monkeys manipulated a joystick to move a virtual arm and reach toward objects on a computer screen. **C**, Experimental task sequence. In the illustrated case, the selection of the right-hand object resulted in reward. The activity of both AT and noise ICMS channels are shown below each box.

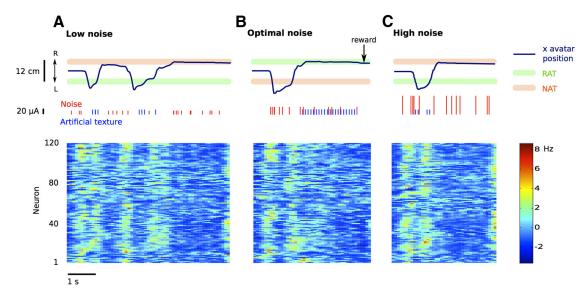


Figure 2. Detection of signal in noise and M1 modulations. Example traces of horizontal movement of the virtual hand for three trials with different noise levels: low (A), optimal (B), and high (C). The colored horizontal bars represent the position of the targets (green, RAT; brown, NAT). In the trials shown in A and C, the monkey incorrectly selected the NAT, whereas in B it correctly picked the RAT and received a reward (indicated by a black arrow). The pulses delivered through both stimulating channels are shown. The RAT is subthreshold. The corresponding neuronal ensemble activity (monkey M, 120 units from M1) is shown in the bottom plots. Note that neuronal modulations reflect the speed and direction of arm movements. The color scale shows normalized firing rate (in Hz).

Materials and Methods

Two rhesus monkeys, M (male) and N (female), were chronically implanted with cortical stainless steel microelectrode arrays in M1 and S1 in both hemispheres (Fig. 1A) (Nicolelis et al., 2003). Neuronal activity was sampled in the arm and hand areas of M1 and S1 in the hemisphere contralateral to the working arm. We used four electrodes of the S1 implants for ICMS delivery. Stimulation was applied to the right hemisphere arm area of S1 in monkey M and the right hemisphere leg area of S1 in monkey N using a custom-built, four-channel, current-controlled stimulator (Hanson et al., 2012). The monkeys were trained to move an

image of the forearm and the hand (animated using MOTIONBUILDER; Autodesk) on a computer screen with a joystick (Fig. 1*B*). Using this virtual hand, they explored virtual objects presented on the screen, which were visually identical, yet differed in their ATs. Each trial started with the monkeys placing the virtual hand over a circular target that appeared at the center of the screen. At this time, ICMS noise of randomly chosen intensity started and continued until the end of the trial. Following the initial central hold, two visual objects appeared on the left and on the right of the screen, equidistant from the center. Then, the monkeys freely explored each of these objects to find the one associated with a periodic

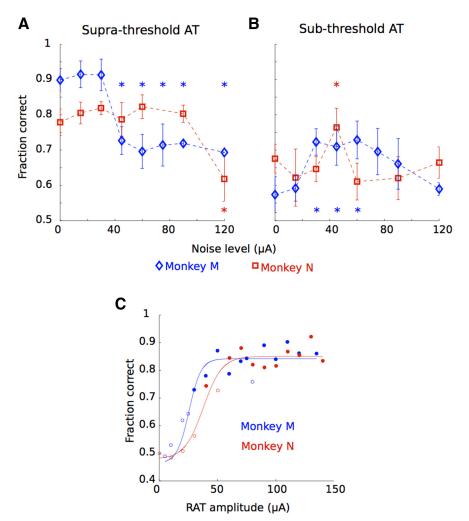


Figure 3. Stochastic facilitation via ICMS. **A**, **B**, Fraction of trials correct per session in terms of the noise amplitude level for monkey M (blue diamonds) and N (red squares) for suprathreshold (**A**) and subthreshold (**B**) RAT. *Significantly different from average performance (p < 0.05 in all cases, randomization test). **C**, Psychometric curves (in terms of ICMS amplitude) for monkeys M (blue) and N (red). Lines, Best sigmoidal fits; open circles, performance at chance level (p > 0.05, Wilcoxon rank sum test).

ICMS pattern, which we call the rewarded artificial texture (RAT). The other object lacked such ICMS feedback [null artificial texture (NAT)]. The RAT consisted of pulse pairs (with 10 ms interpulse interval for monkey M and 5 ms for monkey N) delivered at a frequency of 10 Hz. The monkey selected one of the objects by holding the virtual hand over it for 2 s, and a juice reward was given if they correctly selected the RAT object (Fig. 1C). Each ICMS pulse was a symmetric, biphasic, charge-balanced current waveform delivered through a pair of electrodes implanted 3 mm from each other (i.e., bipolar stimulation). For noise delivery, a different electrode pair was used, implanted 1 mm away from the first pair of electrodes (Fig. 1A). The interpulse interval of such an aperiodic signal was drawn from a gamma distribution with the coefficient of variation (ratio of the standard deviation to the mean) equal to 1 (O'Doherty et al., 2012).

All animal procedures were performed in accordance with the National Research Council's Guide for the Care and Use of Laboratory Animals and were approved by the Duke University Institutional Animal Care and Use Committee.

Results

Monkeys M and N were previously overtrained in a similar active exploration task (O'Doherty et al., 2011), so it took them just one and three training sessions, respectively, to achieve \sim 90% accuracy in the two-target task of this study. No noise was presented at

that point. We then measured the psychometric threshold for RAT detection in the absence of noise for each monkey. Current amplitudes in the range 5–150 μ A were tested. The detection threshold (Gescheider, 1997) was evaluated as the current level at which the proportion of correct responses was 0.75. The psychometric curve (see Fig. 3C) for monkey M was slightly shifted to the left compared with that of monkey N and therefore its threshold was lower (31 vs 45 μ A for monkeys M and N, respectively).

In subsequent sessions, ICMS noise was presented throughout each trial. Noise amplitude for each trial was randomly selected in multiples of 15 μ A, up to 120 μ A. Data with noise were collected for eight and nine daily sessions for monkeys M and N, respectively (monkey M: 370 \pm 60; monkey N: 418 \pm 87 trials per session). For these sessions, the RAT amplitude was either suprathreshold (50 μ A) or subthreshold (20 μ A). Representative trials for a subthreshold stimulus combined with different noise intensities are shown in Figure 2. In these examples, the monkey correctly detected the subthreshold RAT (20 μ A) in the presence of an optimal noise (30 μ A) (Fig. 2B), while failing to do so when the noise amplitude was too low (15 μ A; Fig. 2A) or too high (90 μ A; Fig. 2C). Figure 2 also depicts the neuronal ensemble activity recorded in M1. The firing rates of M1 neurons were clearly modulated in association with the arm movements and reflected movement speed and direction.

Figure 3*A* shows the fraction of correct detection (averaged by session) as a function of noise amplitude for suprathresh-

old RATs (50 μ A). Note that in the absence of noise, monkey M performed better (89.8% correct) than monkey N (77.9% correct), which is consistent with a more rapid saturation of the psychometric curve for monkey M (Fig. 3C). For monkey M, the performance decayed from this high level after noise amplitude exceeded 30 μ A [significantly lower success rate compared with the average for each noise level >30 μ A, p < 0.05, randomization test (Edgington and Onghena, 2007)]. It is possible that we only observed decay simply because the performance was close to saturation from the start and could not improve in principle. However, for monkey N, whose performance level had room for improvement, we observed a similar effect: the performance was resistant to weak noise and started to deteriorate when noise amplitude exceeded 120 μ A (p < 0.001, randomization test).

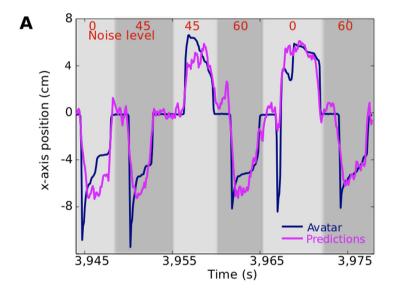
Changes in the monkey's detection performance with noise addition were very different when subthreshold RATs (20 μ A) were used. This dependency was characterized by a nonmonotonic curve with peaks at noise amplitude of 60 and 45 μ A for monkeys M and N, respectively. Thus, for monkey M, elevated performance of ~0.7 was observed for three noise amplitudes: 30, 45 and 60 μ A (p < 0.05 for all points, randomization test). For monkey N, the curve had a single peak of 0.76 at 45 μ A (p <

0.01, randomization test). Therefore, in both monkeys, the addition of a moderate level of ICMS noise to a weak AT stimulus improved the animal's detection performance. Furthermore, neither the presence of ATs nor the addition of noise substantially affected movement-related modulations observed in M1 (Fig. 2), nor did it degrade the performance of a decoding algorithm that extracted the position of the virtual arm from the population activity of M1 neurons [unscented Kalman filter (Li et al., 2009); monkey M: 125 units, monkey N: 65 units; Fig. 4].

Discussion

The present findings demonstrate for the first time SF effects for an artificial tactile sensation. The ability of monkeys to detect weak ICMS signals was clearly improved by the addition of moderate levels of ICMS noise. This result is consistent with previous model and experimental demonstrations of SF in physiological systems (Moss et al., 2004). Interest in SF for neural systems has dramatically increased over the last 20 years, driven by theories claiming that noise may play an important role in normal neural processing (Mc-Donnell and Abbott, 2009; McDonnell and Ward, 2011). Since the presence of moderate amounts of internal noise may be normal and even advantageous for neural systems, several publications proposed implementing noise and SF in artificial systems designed to restore normal functions to people with disabilities, for example in cochlear implants (Morse et al., 2007), assistive devices for postural balance (Harry et al., 2005), and hand sensorimotor tasks (Kurita et al., 2011). Here

we demonstrated the practicality of this idea for prosthetic sensation evoked by ICMS. By adding background noise, we effectively gained control over the detection threshold for an artificial tactile sensation. It has yet to be determined how SF would work in the case of prosthetic sensations evoked by spatiotemporal input patterns delivered through multielectrode implants (Fitzsimmons et al., 2007). For such multichannel systems with stimulation sites distributed over the cortical surface, the superposition of spatiotemporal noise patterns could enhance signal detection in both the temporal and spatial dimensions. Furthermore, SF patterns distributed across corticothalamic loops could be obtained in sensorized neural prosthetics that use thalamic stimulation as an additional channel for artificial somatic sensation (Heming et al., 2010). Although in our present study we did not manipulate the statistics of the noise channel, it would be of great interest in the future to examine a variety of noise-frequency characteristics. If noise contains particular frequencies, SF may accordingly filter frequency components of the input signal. Importantly for practical neuroprosthetic systems, the potential risks of neural damage associated with ICMS may be lower for a SF-based neural prosthetic because of the lower currents injected into the neural tissue (McCreery et al., 2010). In



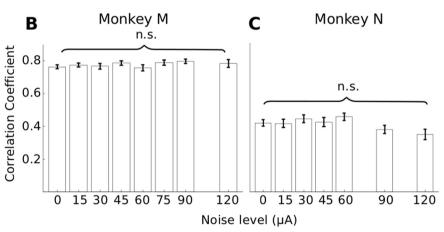


Figure 4. Movement predictions from M1 activity. A, Position of the avatar and off-line predictions from M1 activity for six consecutive trials with different noise amplitudes. B, C, Performance of the decoding algorithm measured as the correlation coefficient in terms of noise level for monkey M (B) and monkey N (C). n.s., No significant difference (one-way ANOVA test, n = 1724, p > 0.57 for monkey M; n = 3421, p > 0.07 for monkey N).

conclusion, we suggest that the incorporation of SF effects in sensorized neural prosthetics may enhance a variety of functional capabilities.

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