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3	Title: Creating a neuroprosthesis for active tactile exploration of textures
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38 Abstract

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Intracortical microstimulation (ICMS) of the primary somatosensory cortex (S1) can 40 produce percepts that mimic somatic sensation and thus has potential as an approach to 41 42 sensorize prosthetic limbs. However, it is not known whether ICMS could recreate active texture exploration—the ability to infer information about object texture by using one's fingertips to scan 43 a surface. Here we show that ICMS of S1 can convey information about the spatial frequencies 44 of invisible virtual gratings through a process of active tactile exploration. Two rhesus monkeys 45 scanned pairs of visually identical screen objects with the fingertip of a hand avatar, controlled 46 via a joystick and later via a brain-machine interface, to find the one with denser virtual gratings. 47 The gratings consisted of evenly spaced ridges that were signaled through ICMS pulses 48 49 generated when the avatar's fingertip crossed each ridge. The monkeys learned to interpret 50 these ICMS patterns evoked by the interplay of their voluntary movements and the virtual textures of each object. Discrimination accuracy across a range of grating densities followed 51 Weber's law of just-noticeable differences (JND), a finding that matches normal cutaneous 52 sensation. Moreover, one monkey developed an active scanning strategy where avatar velocity 53 54 was integrated with the ICMS pulses to interpret the texture information. We propose that this approach could equip upper-limb neuroprostheses with direct access to texture features 55 acquired during active exploration of natural objects. 56

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58 Introduction

59 Sensory neuroprostheses offer the promise of restoring perceptual function to people with impaired sensation ^{1,2}. In such devices, diminished sensory modalities (e.g., hearing ³, 60 vision ^{4,5}, or cutaneous touch ^{6–8}) are reenacted through streams of artificial input to the nervous 61 62 system, typically using electrical stimulation of nerve fibers in the periphery or neurons in the central nervous system. Restored cutaneous touch, in particular, would be of great benefit for 63 the users of upper-limb prostheses, who place a high priority on the ability to perform functions 64 without the necessity to constantly engage visual attention⁹. This could be achieved through the 65 addition of artificial somatosensory channels to the prosthetic device ¹. Such an approach would 66 endow persons suffering from limb loss ^{10–12}, paralysis ^{1,13} or somatosensory deficits with the 67 ability to perform active tactile exploration of their physical environment and aid in dexterous 68 69 object manipulation ^{14–17}.

70 Previously we demonstrated that motor and sensory functions could be simultaneously enacted though a bidirectional neuroprosthetic system, called a brain-machine-brain interface 71 (BMBI)¹⁸. In that demonstration, the active exploration enabled by our BMBI-driven 72 73 neuroprosthesis used a limited and fixed set of ICMS temporal patterns to generate artificial 74 sensory inputs that mimicked the sense of flutter-vibration. However, it remained unclear whether the same approach could generalize to allow the use of natural haptic exploratory 75 procedures, where a person identifies the texture of objects and materials by scanning them 76 with the fingertips. 77

Normal haptic exploration of objects involves several stereotypic procedures, such as static contact for temperature sensation, holding for weight, enclosure for gross shape, pressure for hardness, contour following for exact shape and lateral fingertip motion for texture ¹⁹. Here we developed a neuroprosthetic paradigm for restoring the sensation of fingertip motion against texture. We hypothesized that ICMS pulses generated by exploratory movements over virtual gratings and delivered to primary somatosensory cortex (S1) would allow discrimination of
 texture coarseness.

85 **Results**

86 Active texture encoding

87 Two rhesus monkeys (monkey M and monkey N) were chronically implanted with multielectrode cortical arrays ¹⁸ (Supplementary Fig. S1). These animals explored virtual objects 88 on a computer screen using a realistic upper-limb avatar (Supplementary Fig. S2), which they 89 operated manually with a joystick (Fig.1a) or using a BMI. On each trial, a pair of rectangles 90 91 appeared either on the left or on the right side of the screen. The rectangles were visually identical, but each was associated with an invisible tactile grating whose properties were 92 signaled by charge-balanced ICMS pulses applied to S1 (a region exhibiting left forearm 93 94 receptive fields for monkey M and left lower-limb receptive fields for monkey N). Each grating 95 consisted of evenly spaced vertical ridges, which were invisible to the monkeys. The spatial frequency of the ridges, f, ranged from 0.5 to 4.0 ridges/cm; an untextured object with no ridges 96 (f = 0 ridges/cm) was also presented on some trials. 97

The behavioral task required the monkeys to probe the rectangles with the avatar's 98 99 fingertip, determine which of the two had a higher f, and to hold the avatar over that object for the required interval, 2 s in most cases (Fig. 1b). The artificial sensation was encoded by 100 101 delivering a charge-balanced ICMS pulse each time the avatar fingertip crossed a ridge in a grating. Thus, the pulse-trains of ICMS delivered on any given trial provided an artificial signal 102 that depended on the interplay between the movements of the avatar and the f of the textures of 103 104 the explored objects (Supplementary Movie S1). Movements at a constant velocity across a grating with a given f produced an ICMS pulse train with a constant temporal pulse rate (Fig. 105 106 2a). Movements at a faster velocity across the same grating produced a pulse train with a correspondingly higher pulse rate (Fig. 2b). Irregular movements produced temporally varying 107 ICMS pulse trains (Fig. 2c). The objects' adjacent spacing on the screen encouraged the 108

109 monkeys to rapidly shift the avatar from one object to the other and determine which one had a 110 denser grating. The monkeys were permitted to explore the objects in any sequence and enter each object multiple times, to accumulate evidence, before making the selection. Accordingly, 111 the monkeys could select an object on the first pass (Fig. 2d,e) or employ several explorations 112 113 of individual objects (Fig. 2f) before making a final selection. Prior to these experiments, these monkeys participated in other studies ^{18,20,21} and became proficient in using the joystick and the 114 hand avatar and making decisions using ICMS pulse trains. However, none of the previous 115 experiments employed the particular ICMS encoding rule or the texture scanning paradigm 116 117 presented in the current study.

118 Active texture discrimination

Both monkeys learned the task rapidly, reaching high-performance levels (71% of 119 120 correct trials for monkey N, and 73% for monkey M) after 10 daily sessions of training (Fig. 121 3a,b). The average performance was above chance even in the first training session (64% for monkey N and 56% for monkey M). For monkey M, task difficulty was increased gradually, with 122 a large difference in f introduced early in training, $\Delta f \ge 2$ ridges/cm; $\Delta f < 2$ ridges/cm after 3 123 sessions and the full range from f = 0 to f = 3.5 ridges/cm and a minimum $\Delta f = 0.5$ by the end of 124 the training. The range for monkey N was f = 0 to f = 3.5 ridges/cm at the onset of the training 125 and f = 0 to f = 4 by the end. The minimum difference between textures, Δf , was maintained at 126 0.5 for all sessions. Figure 3 c.d shows the behavioral performance after learning (11 and 12 127 recording sessions for monkeys M and N, respectively). Both monkeys performed better on 128 129 individual trials when presented with larger Δf between the two objects than for smaller Δf , as might be expected. However, we observed an additional scaling of discrimination difficulty that 130 depended on the absolute scale of the spatial frequencies of the objects being compared. More 131 specifically, the psychometric functions for both monkeys were steeper for larger values of Σf , 132 that is, steeper for the larger sum for the two objects being compared (Fig. 4a,b). 133

We quantified this phenomenon by estimating the just noticeable difference (JND), for each presented spatial frequency ²². We calculated, for each spatial frequency, the probability of choosing a second comparison frequency as a function of the unsigned delta between the standard stimulus and the comparison stimulus (Fig. S3). We found that the JND increased proportionally to *f* (Fig. 4c), consistent with the Weber–Fechner law ²³ and Steven's power law ²⁴. The results for monkey M could be described by the linear function JND(f) = 0.47f +1.06 (R² = 0.63); JND(f) = 0.37f + 0.77 for monkey N (R² = 0.95).

There are a number of strategies that the monkeys could have used to compare the 141 textures. One viable option would be to use a consistent velocity when exploring both objects so 142 143 that any variation in ICMS pulse rate between the objects would be due to differences in spatial frequency alone. Further analysis revealed that this was not the case. Indeed, both monkeys 144 used a distribution of speeds to sample the gratings (Fig. 5a) and could perform successful 145 146 discriminations across the majority of their operating range (Fig. 5b)—only having difficulty when 147 moving at very high speeds. Moreover, for the vast majority of trials, the average speeds used 148 to scan the two objects differed, even within the same trial. Monkey M sampled the two objects 149 with the same speed (delta speed < 1 cm/s) on fewer than 3% of trials, a finding that was not 150 explained by the trial outcome (wrong trials: 2.41%, correct trials: 2.82%; Fig. S4). Monkey N 151 used the same scanning speed for each target on only 3.85% of the trials (3.95% of the wrong trials, 3.81% of the correct trials). 152

This variability in arm movements was sufficiently large that, in some cases, the ordinality of spatial frequency of the textures was different from the ordinality of the ICMS pulses rates. An example of one of these apparently paradoxical trials is given in Figure 5c. For this trial, frequency of the right target ($F_R = 3.5$ ridge/cm) was higher than the left ($F_L = 2.5$ ridges/cm), but the actual ICMS pulse rate delivered for the left target was higher than for the right (left: 200.2 Hz versus right: 103.1 Hz). This occurred because a faster avatar speed was used to explore the left target as compared to the right. Despite this, the monkey was able to
 accurately choose the target with the higher spatial frequency in this example.

We found many of these apparently paradoxical trials (n=1231, 12% of all trials) for monkey N. The majority of these cases corresponded to frequency pairs with high FR+FL (Fig. S5). Monkey N's success rate was significantly above chance for these trials (56.1%, P<0.001, one-tailed binomial test; Fig. 5d). There were fewer of these trials for monkey M (n=329, 6% of all trials). For these trials, monkey M's performance did not reach significance (52.03%, P= 0.25, one-tailed binomial test).

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Brain-machine-brain interface with active texture discrimination

Finally, we validated our stimulation paradigm in a closed-loop brain-machine-brain 168 interface (BMBI) with monkey M. For this task, the monkey was allowed to move its arms, but 169 170 the joystick was disconnected; instead the avatar arm-and task performance-was controlled 171 via the decoding of 90 simultaneously recorded right-hemisphere M1 neurons (Fig. 6a). We found that monkey M was able to control the avatar arm to explore the objects with minimal 172 movement of its physical hand as can be seen in the examples shown in Figure 6b. Moreover, 173 when the hand did move, it made smaller movements with lower velocities than the 174 175 simultaneous movements of the cursor during BMI trials (n=63 trials; Fig. 6c), but the monkey could still control the cursor using cortical activity alone (Supplementary Movie S2). The monkey 176 retained the ability to accurately discriminate between the targets using the BMI; consistent with 177 178 the non-BMI task, the monkey was significantly above chance in discriminating targets with Low 179 $\sum f$ (76%, P=0.02, one-sided binomial test), but did not reach significance for medium (65%, P= 0.09) or high $\sum f$ (40%, P = 0.21; Fig. 6d). 180

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182 Discussion

183 We have demonstrated a novel encoding strategy for texture representation using ICMS 184 pulses in somatosensory cortex. Using this new approach, two animals were able to discriminate texture coarseness during active tactile exploration. Importantly, for this task, small variations of arm velocity changed the stimulation frequency; the interpretation of the texture, therefore, may have employed a dynamic integration of ICMS stimulation information with arm proprioception feedback or corollary discharge of motor and sensory cortical regions ²⁵. The apparently paradoxical trials provided evidence for these possibilities: access to the movement command or proprioceptive feedback about the movement is necessary to disambiguate the exafference of the texture from the reafference due to movement.

We observed that both monkeys were better at discriminating textures when the overall spatial frequencies were small, consistent with the Weber-Fechner law ²⁶, a phenomenon reported for numerous sensory modalities ²⁷, including touch ²⁸. Interestingly, this law was previously reported not to hold for the task of discriminating ICMS amplitude in primates ²⁹ and humans ¹³. Our task, in contrast, required discriminating ICMS pulse rates, but, as it also used active exploration we cannot rule out the possibility that some aspect of the effect is due to the motor act itself.

Our tactile encoding scheme was effective for a single channel of independent tactile 199 information-mimicking a single mechanoreceptor localized in the fingertip. This encoding 200 201 scheme most closely resembles the rapidly adapting (RA) afferents of cutaneous somatic sensation ³⁰: each pulse of ICMS was triggered by the intersection of the active zone of the 202 avatar fingertip with a ridge on one of the gratings. However, there may be advantages of 203 modeling a more slowly adapting type-1 (SA1) encoding on some additional channels. We 204 believe that our encoding will be naturally extendable to arrays of mechanosensors embedded 205 in the "skin" of a prosthetic limb, with each sensor connected to a channel of microstimulation in 206 sensory cortex. For example, each feature in an object's tactile microstructure could trigger a 207 pulse-train of ICMS that persists for some finite duration. This type of encoding may allow an 208 intuitive representation of the persistence of object-actuator contact interactions or complex 209 representation of natural textures ³¹. However, a number of open questions remain, such as the 210

optimal timescale or distribution of timescales for adaptation and whether the degree of adaptation must be matched to the properties of the specific neurons being stimulated. Work in primates 6,32 and rats 33 suggests that the plasticity of the brain will allow even a few channels of stimulation to become effective at providing a rich sensory experience, and complex spatiotemporal coding 34 with enough bandwidth to be clinically useful.

In our experiment, monkey N was superior to monkey M in perceiving small differences 216 of texture coarseness. While it is possible that this difference was due to a better 217 comprehension of the task by monkey N, it could also reflect the fact that the stimulation region 218 219 for monkey N was in the leg area while for monkey M it was in the receptive fields of the same arm used to control the joystick. Therefore, it is possible that interference between feedback 220 from natural somatosensory pathways (hand touching the joystick, proprioception) and S1 ICMS 221 222 feedback made interpretation more difficult for monkey M. This indicates that further studies are 223 necessary to determine, among other things, the best target in S1 for delivering ICMS that encodes tactile signals for future clinical neuroprosthesis. While delivering sensory feedback to 224 an ethologically meaningful cortical area is likely important for the subject to assimilate any limb 225 prosthesis as a natural appendage ^{35–37}, the use of different somatosensory regions in the 226 cortex may facilitate the sensory-motor integration and tactile acuity. Therefore, we suggest that 227 it may be necessary to deliver artificial sensory feedback to multiple cortical regions 228 simultaneously to achieve the best performance of such limb prostheses. 229

Recently demonstrated clinical neuroprostheses have used modulation of stimulation amplitude (or equivalently, pulse-width) to encode the perception of pressure, force or position ^{8,10,38,39}. Our approach is complementary—stimulation pulse timing encodes coarse texture—and could be combined with the amplitude encoding approach to convey multimodal percepts of pressure and texture. However some previous animal ⁴⁰ and human ⁴¹ stimulation studies have provided indirect evidence that changes in pulse intensity (amplitude or pulse-width) may be perceptually indistinguishable from changes in pulse rate. Further experiments will be necessary to conclusively determine if this is the case or if there is in fact an extra degree of freedom that
 can be used to convey clinically relevant prosthetic sensations.

Finally, we demonstrated that our encoding strategy could be integrated within a closedloop BMBI task. While the overall performance of the monkey for the BMBI task was lower than during arm-control, the monkey was still able to discriminate the artificial textures. This, along with the simplicity of our ICMS encoding, suggests that this approach could be used to equip clinical upper-limb neuroprostheses with direct access to the tactile features of the natural world.

245 **Online methods**

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.

249 Subjects and Implants

Two adult rhesus macaque monkeys (Macaca mulatta) participated in the experiments 250 (monkeys M and N). Each monkey was implanted with four 96-microwire arrays constructed of 251 insulated stainless steel 304. Each hemisphere received two arrays: one in the upper-limb 252 253 representation area and one in the lower-limb representation area of sensorimotor cortex. These arrays covered both M1 and S1; only microwires implanted in S1 were used for delivering 254 ICMS in study. For the BMI task, we used recordings from the right hemisphere arm arrays as 255 the monkey manipulated the joystick with the left arm. Within each array, microwires were 256 257 grouped in two four-by-fours, uniformly spaced grids each consisting of 16 electrode triplets. The separation between electrode triplets was 1 mm. The electrodes in each triplet had three 258 different lengths, increasing in 300-mm steps. The penetration depth of each triplet was 259 adjusted with a miniature screw. After adjustments during the month following the implantation 260 261 surgery, the depth of the triplets was fixed. The longest electrode in each triplet penetrated to a depth of 2 mm as measured from the cortical surface. 262

263 Task

Each monkey sat in a primate chair, faced a computer screen and grasped a joystick with their left hand. The joystick handle contained an optical sensor to indicate when the monkey released it. The monkeys were trained to manipulate the joystick to control the movements of a left upper-limb primate avatar on the screen ^{18,42}.

Each trial began with a circular target appearing in the center of the screen. The monkeys held an index finger of the avatar within this target for a random delay randomly drawn from a uniform distribution parameterized from 200 to 2000 ms. After this delay, the central target disappeared, and two rectangular object zones appeared on the screen. These appeared either both on the left side or both on the right side of the screen at a distance of 7 cm from the center. Both objects in the pair had the same width, (6 cm). The spacing between the objects was 0.1 cm.

275 Vertical square-wave gratings were superimposed on each of the objects. These gratings, which were not visible to the monkeys, were aligned on the center of each object and 276 were parameterized by spatial frequency, f. When the index finger of the avatar crossed a single 277 ridge in a grating, a pulse of ICMS was delivered to a pair of electrodes implanted in S1 cortex. 278 279 In this way, the pattern of ICMS delivered depended on the velocity of the avatar and the intrinsic spatial frequency of each grating. The microstimulator was serviced at 100 Hz, which 280 meant that for sufficiently fast velocities or high spatial frequencies, it could be possible that 281 more than a single ridge was crossed in a 10 ms interval. If this occurred, we delivered N pulses 282 283 at N*100 Hz, where N was the number of ridges crossed since the last clock cycle. This operation delivered the correct number of pulses at the correct rate, in expectation, at the cost 284 of up to 10 ms of latency. 285

286 Symmetric, biphasic, charge-balanced, cathode-leading ICMS pulses were delivered in a 287 bipolar fashion across pairs of microwires. The channels selected had clear sensory receptive 288 fields in the left forearm (monkey M: two pairs of microwires) or left lower limb (monkey N: one pair of microwires). For monkey M, the cathodic and anodic phases of stimulation had a pulse width of 105 μ s; for monkey N, the pulse phases were each 200 μ s. The cathodic and anodic of the stimulation waveforms were separated by a 25 μ s interphase interval. The pulse amplitudes were set to the minimal effective current, as found through psychometric measurements separately for each monkey ⁴³.

294 Monkeys received a reward for selecting the object from the pair with the higher spatial 295 frequency, *f*, drawn from:

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$f \in \{0, 0.5, 1.0, 1.5, 2.0, 2.5, 3.0, 3.5, 4.0\}$ ridges/cm

with the constraint that both objects did not share the same f on a single trial. Monkey M did not discriminate the gratings as reliably and so was not presented any gratings with the highest spatial frequency, 4.0 ridges/cm. The monkeys indicated their choice by holding the avatar over one of the objects for the hold interval (2 seconds for the hand control and 1 second for the BMBI task). Selecting the object with the higher f triggered the delivery of a fruit juice reward; selecting the object with lower f ended the trial without reward.

The objects could be explored in any sequence. Moreover, objects could be re-explored and re-compared multiple times in a trial. However, the avatar had to pass over both objects at least once per trial. Trials for which only a single object was explored were terminated without reward, even if the correct object was ultimately selected. Trials for which the monkey released the joystick handle at any time, selected the wrong object, made a selection without exploring both objects, or held the avatar outside of either of the objects for 10 s, resulted in the termination of a trial and penalty interval of 2 s for monkey M and 2.5 s for monkey N.

We employed correction trials. This meant that after an incorrect trial, the next one repeated with the same object locations and object-frequency identities. These correction trials were used to keep the monkeys motivated and to prevent them from acquiring systematic

biases. As the rewarded object was known to the monkeys for correction trials, we excludedthese trials from all analyses.

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316 BMI decoding

A 10th-order Unscented Kalman filter (UKF) was used for BMI predictions, using methods we previously described ^{18,44}. The filter parameters were fit using the hand movements made while the task was performed using a joystick. The monkey was permitted to continue moving the joystick, but was only rewarded for target selections made with the brain-controlled cursor.

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434	
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437	
438	Materials & Correspondence. The custom code used for the experiments and the data that
439	support the findings of this study is available from the corresponding author upon reasonable
440	request. Requests should be addressed to nicoleli@neuro.duke.edu
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445 **Figure Legends**

446

Fig. 1. The artificial texture paradigm. (a) A monkey is seated before a display on which an 447 448 avatar arm and two identical objects are projected. Artificial tactile feedback about the virtual gratings associated with each object is delivered to populations of S1 neurons via temporal 449 patterns of ICMS as the monkey actively scans each object. (b) Trials commenced with a 450 random delay interval (1) when the monkey held the index finger of the avatar in the center of 451 the screen. Next, was the exploration interval (2). Two rectangular objects appeared, and the 452 453 monkey scanned these objects with the index finger of the avatar hand. Each object had an associated virtual grating of vertical lines, which were invisible to the monkey. A pulse of ICMS 454 was delivered to a pair of electrodes in S1 with each crossing of the avatar index finger over a 455 456 line in one of the gratings. The trial was completed when the monkey indicated its selection (3) by holding the avatar hand over one of the objects for a hold interval. The reward was delivered 457 if the monkey selected the object with the higher virtual grating frequency (inset); selecting the 458 object with the lower grating frequency ended the trial without reward. 459

460

Fig. 2. The precise temporal pattern of ICMS delivered on any trial depended both on the 461 intrinsic spatial frequency of each object's virtual grating as well as the velocity with 462 which the monkey scanned each object. For a grating with a given spatial frequency, slow 463 scanning (a) would produce a lower ICMS pulse rate than faster scanning (b). Irregular 464 scanning (c) of a grating produced irregular ICMS pulse trains. All other features of the pulse 465 train (e.g., current amplitude and pulse width) were fixed. (d-f). Examples of trials for three 466 values of Δf : (d) 3.5 (4.0 vs 0.5) ridges/cm, (e) 2.0 (2.0 vs 4.0) ridges/cm, and (f) 0.5 (2.0 vs 1.5) 467 ridges/cm, respectively. Traces indicate the x-component of the avatar position (solid lines) and 468 velocity (dashed lines). Gray rectangles indicate the position and horizontal dimension of the 469

470 objects. Red vertical lines indicate single pulses of ICMS. Trials started with a randomized hold-

time (200-2000 ms); a Go cue informed the monkey of the beginning of the exploration interval.

472

473 Fig. 3. Monkeys discriminated spatial gratings based on self-generated temporal ICMS.

474 (a-d) Percentage of trials for which the monkey chose the right-most object, parameterized by the spatial frequencies of the right and left objects, for both monkeys. (a,b) The success rate at 475 the first session, and after three and after 10 sessions of training, is reported in parenthesis, and 476 the average percentage of trails for which the right object was chosen when $f_R > f_L$ and when 477 $f_L > f_R$, reported in yellow and blue, respectively. (c,d) The performance for all sessions (11 478 sessions for monkey N, n=10412 and 12 for monkey M, n=5828); monkey M was not presented 479 gratings with 4.0 ridges/cm. Asterisks indicate frequency-pair combinations which were 480 481 discriminated significantly differently than chance (P<0.05, two-sided binomial test).

482

Fig. 4. Psychometrics analysis of artificial texture discrimination. (a,b) Discrimination of 483 spatial gratings obeys Weber's scaling for (a) monkey N and (b) monkey M. Each point 484 represents the percentage of trials for which the monkey chose the right-most object, 485 486 parameterized by the difference in spatial frequencies for a pair of objects (Δf , f_R - f_L) and the sum 487 of the spatial frequencies (f_R+f_l) for low (less than 2.5 ridges/cm, circles), mid (between 2.5 and 5 ridges/cm, diamonds) and high (greater than 5 ridges/cm, triangles) sums. Filled symbols 488 indicate discrimination significantly different than chance (P<0.05, two-sided binomial test). 489 Error-bars indicate 95% confidence intervals. Curves are the sigmoid lines of best fit. (c) Just 490 noticeable differences (JNDs) for monkey M (diamonds) and N (circles), as a function of the 491 492 standard frequency (detail of JND calculation for each standard frequency is shown on Figure S3: for monkey M JNDs for f =1 and f= 1.5 were undefined). Linear fits, the corresponding 493 function and R^2 for each graph. 494

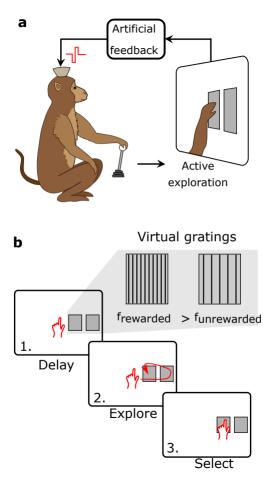
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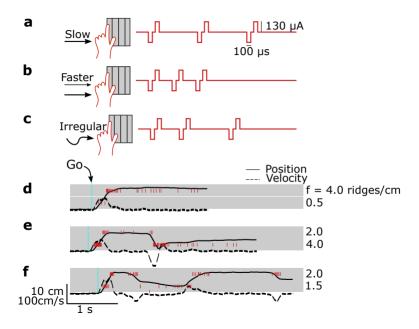
496 Fig. 5. Texture perception and arm movement. (a) Distribution of per-trial RMS exploration speeds. (b) Percentage of trials performed correctly as a function per-trial RMS speed for 497 monkey M and monkey N. Curves are 4th order polynomial fits. Filled symbols indicate 498 499 discrimination significantly different than chance (P<0.05, two-sided binomial test). (c) An 500 example of a paradoxical trial with monkey N. First two graphs indicate the x-component of the avatar velocity and the x-position. Gray rectangles indicate the position and horizontal 501 dimension of the objects; their corresponding spatial frequencies were 2.5 and 3.5 ridges/cm, 502 respectively. Vertical red lines indicate single pulses of ICMS. ICMS pulse rate were calculated 503 for bursts of stimulation (a burst of stimulation was considered when the velocity magnitude was 504 maintained above 10 [cm/s]). (d) Success rate of the paradoxical trials. The chance level is 505 reported with a black dashed line. Error-bars indicate 95% confidence intervals (one-sided 506 507 binomial test).

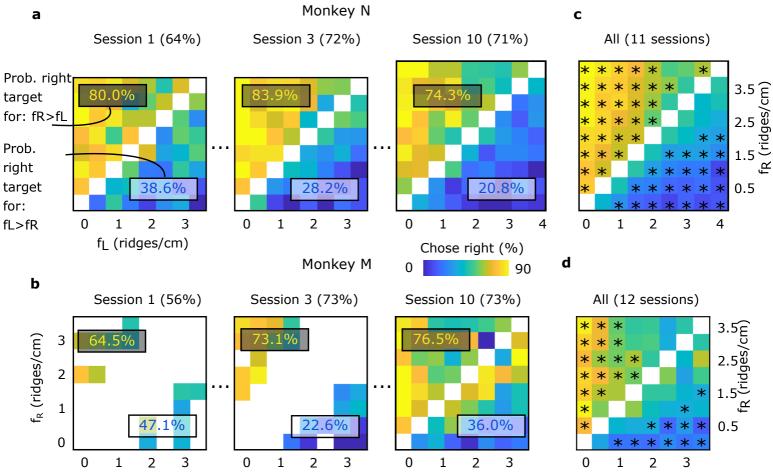
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Fig. 6. BMI results (a) Same experimental paradigm as in Fig 1a, expect that the control of the 509 avatar arm was done via decoding of motor intention from monkey's motor cortex. The monkey 510 has to hold the joystick during the task, was allowed to move the arm, but the joystick was 511 512 disconnected (b) Examples of two BMI trials and corresponding raster plots. Blue dashed lines report the x projection of the brain-controlled cursor (BMI) and solid black line the monkey's 513 hand movement. ICMS pulses are shown with red vertical lines. Vertical dashed cyan line is the 514 end of the hold time (or onset of exploration), and solid green line is the end of the trial and 515 reward. Raster plots for the trials are grouped between 90 neurons in right hemisphere motor 516 cortex area (R-M1), 47 neurons in right hemisphere sensory area (R-S1) and five neurons in the 517 left hemisphere sensory area. (c) The distributions of velocity for the BMI controlled cursor 518 519 (blue) distribution of the monkey's hand movements (orange). The hand movement was 520 measured via the joystick movement (using the same gain as for hand control trials) and the trial was aborted if the monkey released the joystick handle. (d) Percentage success for BMI 521

- 522 controlled trials, parameterized by the sum of the spatial frequencies (f_R+f_L) for low (less than
- 523 2.5 ridges/cm), mid (between 2.5 and 5 ridges/cm) and high (greater than 5 ridges/cm) sums.
- 524 Error-bars indicate 95% confidence intervals. Filled symbols are statistically different than
- 525 chance (P<0.05, one-sided binomial test).
- 526







fL (ridges/cm)

