BRAIN-MACHINE INTERFACE

Mikhail Lebedev

Duke University, Durham, North Carolina, U.S.A.

Keywords: brain, cortex, decoding, interface, microstimulation, multielectrode, neurophysiology, neuroprosthetic, paralysis, plasticity, primate, prosthetic, robot

Contents

- 1. General Introduction
- 2. History of Research and Commercialization
- 3. Information Encoding in the Brain
- 4. Motor BMIs
- 5. Neuronal Ensembles and Large-Scale Recordings
- 6. BMI for Reaching and Grasping
- 7. Decoding Algorithms
- 8. Neuronal Plasticity
- 9. Noninvasive BMIs
- 10. BMI for Walking
- 11. Sensory BMIs
- 12. Bidirectional BMIs
- 13. Conclusions and Perspectives
- Related Chapters

Glossary

Bibliography

Biographical Sketches

Summary

Brain-machine interfaces (BMIs) strive to restore neural functions to disabled individuals by linking brain circuitry to artificial devices, such as computers, prosthetic limbs, wheelchairs and communication systems. The BMI field has advanced considerably during the last two decades, driven by the developments in neural recording methods, computer science, robotic engineering and medical technology. BMIs are often classified by their function as sensory, motor or sensorimotor (bidirectional). Motor BMIs extract motor commands from brain signals and redirect them to artificial actuators. Sensory BMIs aim to restore sensory functions - hearing, vision, sense of touch and limb position – by interfacing neural structures with artificial sensors. Sensorimotor BMIs combine sensory and motor components. Additionally, cognitive BMIs have been proposed to decode higher-order brain signals, such as decisions, thoughts and memories. Depending on the degree of interference with the biological tissue, BMIs are classified as noninvasive or invasive. Noninvasive BMIs, for example those utilizing electroencephalography (EEG), are safe to use, but their bandwidth is limited. Invasive BMIs employ electrodes implanted in the brain to get access to highly informative neural signals produced by single neurons and their populations. Many decoding algorithms have been proposed for extraction of information from brain activity and its utilization in BMIs. In addition to clinical goals, BMIs provide insights on fundamental brain mechanisms.

1. Introduction

1.1. Neural Control and When Things Go Wrong

Motor movements are essential for the interaction of living organisms with each other and with the external world. Indeed, all forms of mental activity – from relatively simple to highly sophisticated – are eventually expressed through muscle contractions and relaxations. Muscles move our limbs, rotate our eyes, produce facial expressions, and generate speech. Properly controlled muscle activity is essential not only for motor behaviors, but also for sensory functions. We actively seek sensations: reach toward objects with our arms; touch and grasp these objects to appreciate their texture and shape. As the body as a whole and its individual parts move, their displacements are monitored by numerous sensory receptors located in the skin, muscles, tendons and joints, as well as by the vestibular apparatus and vision. These continuous streams of sensory and motor information are processed by multiple neural structures richly interconnected with each other. We are consciously unaware of the majority of details of this immense neural processing and take for granted that we can effortlessly perform such complex tasks as maintenance of balance, bipedal walking, dexterous hand movements, speech and many others.

Unfortunately, neural trauma, disease or limb loss may seriously disrupt normal physiological functions. Destruction of just a few millimeters of nervous tissue may leave a person unable to move and feel. Spinal cord injury (SCI) breaks communication pathways between the brain and the spinal cord and produces devastating sensorimotor deficits, often a complete paralysis of large portions of the body. Neurological stroke can entail profound motor and sensory deficits. In Parkinson's disease, degeneration of dopamine neurons in substantial nigra pars compacta – a relatively small subcortical nucleus – dramatically damages sensorimotor and cognitive functions. Amyotrophic lateral sclerosis (ALS), a motor neuron disease, results in paralysis, muscle atrophy and eventually death.

Strikingly, higher brain functions often remain intact in many of these dire neurological conditions. Thus, in SCI and locked-in syndrome, patients are unable to produce muscle contractions, but remain mentally able and awake.

Currently, there is no cure for many devastating neurological conditions, such as SCI and ALS. Millions of paralyzed patients are bound to their beds or wheelchairs for the rest of their life. The development of efficient treatments for neurological trauma and disease is clearly one of the most important and difficult challenges for the medical science today.

1.2. Connecting the Brain to Machines

Brain machine interfaces (BMIs) represent an ambitious attempt to revolutionize treatment of paralysis and other neurological conditions. BMI is an artificial system that

enables communication between the brain and artificial devices, such as computers, limb prostheses and neural stimulators (Lebedev and Nicolelis 2006; Nicolelis and Lebedev 2009; Nicolelis 2011; Schwartz et al. 2006; McFarland et al. 2006; Hatsopoulos and Donoghue 2009) (Figure 1). Therapeutic BMIs strive to bypass the site of neural damage and to establish a direct functional connection between an intact brain area and an assistive device. For example, it has been proposed that SCI patients may be able to regain motor function if they are aided with a BMI that extracts signals from the motor cortex and directs these signals to robotic limbs, exoskeletons, or functional electrical stimulation (FES) devices connected to paralyzed muscles (Lebedev and Nicolelis 2006; Wolpaw et al. 2002). In addition to medical applications, BMIs can be employed by healthy people to enhance certain neural and physiological functions.

In addition to the term "BMI", systems interfacing neural tissue with external devices are called brain-computer interfaces (BCIs), mind-machine interfaces, neural interfaces, brain implants or neural prostheses. Although this terminology is commonly used interchangeably, some authors make strict distinction between specific BMI subtypes. In this chapter the term "BMI" is used in its most generic meaning.



Figure 1. Brain-machine interface for reaching and grasping. Neuronal ensemble activity was recorded in multiple cortical areas in a rhesus monkey and translated into reaching and grasping movements performed by a robotic arm. The experimental setup included the data acquisition system, the computer running BMI decoders, the robot arm, and the visual display which provided feedback of the robot movements. Adapted from Carmena et al. (2003)

BMIs introduce artificial components into neural circuitry: sensors for sampling neural signals, electronic chips that decode and transform neural activity, neural stimulation devices, robotic limbs with sensors of touch and position, wireless transmitters, and other components. In science fiction, biological organisms that receive artificial parts are called cyborgs. Whereas BMI-based cyborgs, as envisioned by futuristic writers, may emerge in the future, current BMI research considers medical treatment as the priority and major practical goal. Many branches of this clinically oriented research, particularly research on invasive BMIs, are still at the stage of animal experiments or preliminary human trials. Considerable effort will be required in these areas to develop

fully functional clinical neural prostheses. The major challenges that hinder the progress in clinical BMIs include the need to improve neural recording methods, problems of biocompatibility, the task of making BMIs fully implantable, development of advanced BMI decoders, incorporation of artificial sensation in BMI systems, and engineering of advanced robotic prostheses (Lebedev and Nicolelis 2006). One notable exception is the cochlear implant, which has already entered the clinical world and has helped hundreds of thousands patients to regain hearing (Shannon 2012; Wilson and Dorman 2008)).

In addition to medical applications, BMIs have emerged that are intended for healthy people. These are, for example, BMI devices for computer gaming that allow users to play simply by thinking instead of using a joystick or a keyboard (Tangermann et al. 2009). Additionally, BMIs can provide useful biofeedback of neural signals that people cannot perceive through their normal senses. For example, a safety system for a long-distance driver continuously samples encephalographic (EEG) activity and issues a warning if signs of drowsiness are detected (Lin et al. 2010). In the future, consumer BMIs may allow humans to exceed many of normal abilities: computational power, accuracy, consistency, reaction time and physical strength.

BMI research has experienced a spectacular growth since the 1990s, driven by the progress in multichannel neural recordings, computer technologies and robotic engineering. Many ideas previously entertained only by science fiction are becoming a reality.

1.3. Ethical Considerations and Cognitive BMIs

The prospect of BMI systems being able to reproduce virtually any brain function brings up a number of philosophical and ethical issues (Farah 2002; Vlek et al. 2012). Is it ethical to intrude into a person's mind with a BMI? Is there a danger that BMIs may interfere with the representation of self and free will? Even though many of these questions seem far-fetched, research has already started on BMIs that extract higher-order cognitive signals from brain activity, such as decisions (Andersen et al. 2010) and memories (Berger et al. 2005).

1.4. BMI Types by Function

Currently, the major focus of BMI research is on systems that handle motor and sensory signals. These BMIs hold promise to provide practical solutions for restoration of vital functions to people with disabilities. BMI systems of this type are often classified as: (i) motor, (ii) sensory, or (iii) bidirectional (sensorimotor). This classification resembles a simplified description of the layout of the nervous system as an arrangement of sensory areas, motor areas and their interconnections. In actuality, there is no clear-cut separation between sensory and motor areas in the brain. For instance, cortical areas – even the ones called primary motor and primary sensory – process both motor and sensory information and therefore are best described as sensorimotor (Lilly 1956; Evarts 1973). It is quite possible that with the advancement of the BMI field, BMIs will be incorporating sensorimotor modules instead of segregating motor and sensory processing. Recently demonstrated bidirectional BMIs represent the first step in this development (O'Doherty et al. 2011).

1.5. Invasive and Noninvasive BMIs

Safety is an important consideration when choosing a BMI system. The safest, nonivasive BMIs, utilize sensors (e.g., EEG electrodes) that may come in contact with the skin, but do not penetrate the body. Although such systems are safe to use, their information bandwidth is limited, often resulting in insufficient speed and accuracy of performance (Lebedev and Nicolelis 2006; Wolpaw et al. 2002).

Nonivasive recording methods, such as EEG, utilize weak neural signals detected at a distance from their source. These recordings often have low spatial resolution (the ability to discriminate signals from nearby brain sites) and/or low temporal resolution (the ability to detect rapid neural modulations), but they can be implemented much more easily than invasive recordings. Consequently, noninvasive BMIs have been extensively studied in humans. Many practical noninvasive systems have been developed, such as BMIs for communication, prosthetic control and wheelchair navigation (Galán et al. 2008; Muller-Putz and Pfurtscheller 2008; Nicolas-Alonso and Gomez-Gil 2012; Sellers et al. 2010).

Invasive BMIs hold promise to achieve a much higher bandwidth compared to noninvasive systems. An invasive surgical procedure is required to bring recording sensors close to brain neurons, the signal source. Extracellular activity or single neurons is usually recorded with microwires (typically 10-50 microns in diameter) implanted in the brain (Lebedev and Nicolelis 2006; Nicolelis and Lebedev 2009). In addition to single-unit recordings, microwires can be also used to record local field potentials (LFPs), which represent combined spiking activity and dendritic potentials of many neurons.

Much of invasive BMI research has been conducted in experimental animals (rodents and primates). Human clinical research on invasive BMIs has recently started, with promising results (Hanson et al. 2012; Hochberg et al. 2012; Collinger et al. 2013). Several issues still hinder the acceptance of invasive BMIs in clinic, most importantly the need to achieve reliability and long-term performance of invasive implants and the need to transition from tethered to wireless recordings (Lebedev and Nicolelis 2006).

Electrocorticography (ECoG) is a safer alternative to implanted microwires. This is a minimally invasive recording method that utilizes electrodes placed on the brain surface. Craniotomy is required, but the risk of neural damage is decreased. ECoG recordings cannot resolve discharges of single neurons, but they work well to detect synchronous electrical activity of large populations of neurons.

2. History of Research and Commercialization

2.1. The Birth of BMI Field

The first experiments in which multiple electrodes were implanted in the brain date back to the 1950s when John Lilly implanted 25 to 610 electrodes in monkey cortex at intervals from one to two millimeters apart. He then applied electrical stimulation through these electrodes to elicit movements (Lilly 1956). Lilly observed that movements occurred even when he stimulated cortical areas presumed to be purely sensory. He concluded that each cortical area was best characterized as sensorimotor rather than purely sensory or purely motor.

In the 1960s and 1970s, David Nowlis, Joe Kamiya, Abraham Black, Maurice Sterman and their colleagues experimented with EEG as a source of biofeedback that enabled subjects (both animals and humans) to gain control over their own brain rhythms (Lebedev and Nicolelis 2006).

In 1963, Grey Walter conducted a study which can be considered the first demonstration of a real-time BMI (Dennett 1992). He implanted electrodes in the motor cortex of patients undergoing neural surgery. The patients advanced a slide projector by pushing a button with their hands. These voluntary button presses were preceded by readiness potentials in the motor cortex. To create a direct link between the brain and the projector, Walter disconnected the button and made cortical potentials advance the slides. The patients continued to press the disconnected button, but the control signal came directly from the brain. Remarkably, their direct cortical control often worked before they initiated hand movements.

About the same time, a team of National Institutes of Health (NIH) scientists led by Karl Frank formulated the goal of building neurally controlled prosthetic devices. Frank wrote, "We will be engaged in the development of principles and techniques by which information from the nervous system can be used to control external devices such as prosthetic devices, communications equipment, teleoperators ... and ultimately perhaps even computers" (Frank 1968). The NIH team was able to achieve promising results. In 1970, team members Humphrey, Schmidt and Thompson simultaneously recorded 3-8 neurons with five electrodes inserted in the motor cortex of monkeys performing wrist flexions and extensions (Humphrey et al. 1970). The recordings were stored on tape. In an offline analysis of these data, the researchers were able to extract movement traces from the neuronal rates using multiple linear regression. A decade later, Schmidt recorded from monkey cortex with a 12-electrode array that stayed implanted for 37 months (Schmidt 1980). Schmidt's monkeys learned to move a cursor on a LED display by modulating their cortical activity.

In parallel with this work, pioneering studies on volitional control of single cortical neurons were conducted by Ebernhard Fetz at the University of Washington. His monkeys voluntarily modulated the activity of single cortical neurons in order to attain a particular firing rate (Fetz 1969).

Research on sensory BMIs started at about the same time as the work on motor BMIs. The development of cochlear implants was especially successful (Wilson and Dorman 2008). This work was pioneered in 1957 by Djourno and Eyriès who developed an implant that applied single-channel stimulation to the auditory nerve. The stimulation frequency was up to 1 kHz, and their patient eventually was able to detect pitch differences and recognize words. The first multichannel cochlear stimulator was developed by Blair Simmons in 1964. In the 1970s, William House and Jack Urban introduced a carrier frequency of 16 kHz to the stimulation signal. Their work eventually resulted in clinical device that received FDA approval. Additionally, Robin

Michelson's work contributed to improved implantation methods. Over several decades, cochlear implants have improved and have become widely accepted in clinic. More than 200,000 people have been implanted worldwide with these devices.

Also in the late 1960s – early 1970s, research has started on sensory BMI for restoration of vision. In these early studies conducted by the groups of Brindley and Dobelle (Brindley and Lewin 1972; Dobelle et al., 1974), electrical stimulation was applied through electrodes placed on the surface of the visual cortex in totally blind individuals. The subjects reported phosphens, i.e. appearances of light spots in their visual field. Moreover, they were able to recognize simple patterns and letters composed of such phosphens. These pioneering studies demonstrated the feasibility of a visual prosthesis for the blind.

2.2. Rapid Development and Key Players

In the late 1990s – early 2000s BMI research markedly accelerated, facilitated by progress in multielectrode recording methods and computer technologies. Several researchers became notable players in the BMI field.

Miguel Nicolelis, John Chapin and their colleagues at Hahnemann University pioneered an invasive BMI that converted extracellular activity of neuronal populations recorded in rat cortex and thalamus into one-dimensional movements of a robot (Chapin et al. 1999).

Nicolelis then moved to Duke University where he started BMI research in nonhuman primates. He and his colleagues pioneered several BMIs that utilized large-scale neuronal activity recorded from multiple cortical areas to control external actuators. Aided by these BMIs, monkeys learned to control robot arms that performed reaching movements (Wessberg et al. 2000) and reaching and grasping movements (Carmena et al. 2003; Lebedev et al. 2005). Nicolelis and his colleagues also pioneered cortical microstimulation as a method to provide artificial tactile feedback for a BMI that controlled arm reaching (O'Doherty et al. 2011). Furthermore, the Nicolelis laboratory extended their BMI approach to bipedal locomotion (Fitzsimmons et al. 2009).

Philip Kennedy is another prominent figure in the BMI field. He and his colleagues implanted an ALS patient with a neurotrophic electrode that induces growth of myelinated fibers into the recording tip. The patient was able to achieve an on/off control with neural signals (Kennedy and Bakay 1998).

John Donoghue heads a BMI research group at Brown University. The group conducts invasive-BMI experiments in monkeys and in human patients. They were the first to implant paralyzed patients with multielectrode arrays in the motor cortex. These human studies demonstrated real-time BMI control of a computer cursor (Hochberg et al. 2006) and a robotic manipulator (Hochberg et al. 2012).

Andrew Schwartz, first at Arizona State University and then at the University of Pittsburg, developed invasive BMIs for the control of three-dimensional movements of

cursors (Taylor et al. 2002) and robots (Velliste et al. 2009). His laboratory also develops invasive BMIs for humans (Collinger et al. 2013).

The other notable players in the invasive BMI research are Richard Andersen at California Institute of Technology, Krishna Shenoy at Stanford University and Elon Vaadia at the University of Jerusalem.

In parallel with the development of invasive BMIs, noninvasive BMIs (often called BCIs) achieved impressive progress since the late 1990s. Many practical applications have been developed for assisting disabled patients in using a computer, controlling a wheelchair and restoring mobility of paralyzed limbs. Among the leading researchers in this field are Niels Birbaumer (University of Tübingen), Gert Pfurtscheller (Graz University of Technology), Theresa Vaughan (Wadsworth Center), Klaus-Robert Müller (Fraunhofer Institute for Intelligent Analysis and Information Systems), Gerwin Schalk (Wadsworth Center), Christa Neuper (University of Graz), Andrea Kübler (Eberhard-Karls-University), Jonathan Walpaw (Wadsworth Center) and Jose Millan (EPFL).

2.3. Commercialization

Several companies have emerged that commercialize BMI technology. Philip Kennedy founded Neural Signals, which is a research company focused on the development of communication devices for paralyzed patients. John Donoghue and his colleagues founded Cyberkinetics, a company that developed BMI components and marketed recording equipment. Cyberkinetics was then sold to Braingate. William Dobelle founded Avery Biomedical Devices to develop visual implants, but this work slowed down after his death. Several companies develop EEG-based BMIs: Guger Technologies, Interactive Productline, NeuroSky, Emotiv Systems, Starlab and OCZ Technology. Two major manufactures of cochlear implants are Cochlear Corporation and Advanced Bionics Corporation.

3. Information Encoding in the Brain

3.1. Factors that Allow Decoding of Neural Signals

Notwithstanding many remarkable advances made by neuroscientists, we still have a poor understanding of brain computations that underlie motor control, sensory processing and cognition. Luckily for BMI scientists, this poor understanding does not prevent them from trying and succeeding in extraction of useful information from the brain.

Numerous neurophysiological studies have shown that individual neurons and their populations modulate their discharge rates when the brain processes information. Even though the exact functional role of these neural activities is unknown in many cases, BMI researchers still can develop algorithms that discover correlations between neural signals and behavioral parameters of interest and employ these correlations to extract behavioral parameters from brain activity.

TO ACCESS ALL THE 36 **PAGES** OF THIS CHAPTER, Visit: <u>http://www.eolss.net/Eolss-sampleAllChapter.aspx</u>

Bibliography

Andersen R.A., Hwang E.J., Mulliken G.H. (2010) Cognitive neural prosthetics. *Annu. Rev. Psychol.* 61: 169-190: C1-3. [This review discusses the idea of extracting cognitive signals from the brain and using them to control a brain-machine interface]

Bach-y-Rita P., Kercel W. (2003) Sensory substitution and the human-machine interface. *Trends Cogn. Sci.*, 7:541-546. [A review of sensory substitution systems for restoration of vision and balance]

Barton J.J. (2011) Disorder of higher visual function. *Curr. Opin. Neurol.* 24: 1-5. [This review discusses studies on blindsight, the ability for visual processing in the absence visual areas responsible for visual awareness. The review also discusses studies on ventral stream syndromes, which occur after lesions to medial occipitotemporal structures, and dorsal stream syndromes following bilateral occipitoparietal lesions]

Berger T.W., Ahuja A., Courellis S.H., Deadwyler S.A., Erinjippurath G., Gerhardt G.A., Gholmieh G., Granacki J.J., Hampson R., Hsaio M.C., LaCoss J., Marmarelis V.Z., Nasiatka P., Srinivasan V., Song D., Tanguay A.R., Wills J. (2005) Restoring lost cognitive function. *IEEE Eng. Med. Biol. Mag.*, 24: 30-44. [A review of BMI technologies that allow to replace a part of hippocampal circuitry with a neural prosthesis. Such implant is able to restore memory function]

Birbaumer N., Ghanayim N., Hinterberger T., Iversen I., Kotchoubey B., Kübler A., Perelmouter J., Taub E., Flor H. (1999) A spelling device for the paralysed. *Nature*, 398: 297-298. [A pioneering study where a communication channel for 'locked in' patients was established using a brain-computer interface based on slow cortical potentials]

Birbaumer N., Murguialday A.R., Cohen L. (2008) Brain-computer interface in paralysis. *Curr. Opin Neurol.*, 21: 634-638. [A review of progress and challenges in brain-computer interfaces that restore function to patients with 'locked-in' syndrome]

Brindley G.S., Lewin W.S. (1968) The sensations produced by electrical stimulation of the visual cortex. *J. Physiol.*, 196: 479-493. [A pioneering study on electrical stimulation of the visual cortex in totally blind individuals. A 52-year-old blind patient was implanted with an array of surface electrodes placed over the occipital pole of the right cerebral hemisphere. The electrodes were connected to an array of radio receivers, which were used to trigger sensations of light (phosphenes) in different parts of left half of the visual field. By stimulation through several electrodes simultaneously, the patient could be caused to see predictable simple patterns]

Chase S.M., Kass R.E., Schwartz A.B. (2012) Behavioral and neural correlates of visuomotor adaptation observed through a brain-computer interface in primary motor cortex. *J. Neurophysiol.*, 108: 624-644. [A rotational transformation was applied to a subpopulation of neurons engaged in a brain-machine interface. This perturbation resulted in adaptations that occurred in both the entire population and the local neuronal subset responsible for the error]

Chapin J.K., Moxon K.A., Markowitz R.S., Nicolelis M.A. (1999) Real-time control of a robot arm using simultaneously recorded neurons in the motor cortex. *Nat. Neurosci.* 2: 664-670. [A pioneering study of real-time control of a robot by neuronal ensembles recorded in rat motor cortex and ventralateral thalamus]

Cheron G., Duvinage M., De Saedeleer C., Castermans T., Bengoetxea A., Petieau M., Seetharaman K., Hoellinger T., Dan B., Dutoit T., Sylos Labini F., Lacquaniti F., Ivanenko Y. (2012) From spinal central pattern generators to cortical network: integrated BCI for walking rehabilitation. *Neural Plast.*, 2012:375148. [This paper critically investigates different approaches to prosthetics for the restoration of bipedal walking: EEGs, upper limb EMGs, or a hybrid system. Mindwalker project is introduced, as well]

Cincotti F., Mattia D., Aloise F., Bufalari S., Astolfi L., De Vico Fallani F., Tocci A., Bianchi L., Marciani M.G., Gao S., Millan J., Babiloni F. (2008) High-resolution EEG techniques for brain-computer interface applications. *J. Neurosci. Methods*, 167: 31-42. [In this study, high-resolution electroencephalographic techniques have been shown to improve BMI decoding using EEG pre-processing that removed spatial correlation introduced by current conduction. Subjects utilized this BMI to self modulate sensorimotor EEG rhythms, related to the imagination of limb movements]

Collinger J.L., Wodlinger B., Downey J.E., Wang W., Tyler-Kabara E.C., Weber D.J., McMorland A.J., Velliste M., Boninger M.L., Schwartz A.B. (2013) High-performance neuroprosthetic control by an individual with tetraplegia. *Lancet*, 381: 557-564. [An individual with tetraplegia was implanted with cortical multielectrode array that recorded close to two hundred neurons simultaneously. With these neural signals, the patient was able to operate a BMI that controlled a sophisticated robotic arm]

Cordo P.J., Gurfinkel V.S. (2004) Motor coordination can be fully understood only by studying complex movements. *Prog. Brain Res.*, 143: 29-38. [This review uses the sit-up movement to illustrate the complexity and hierarchical control of coordination in movements that involve many muscles, joints and degrees of freedom]

Courtine G., Gerasimenko Y., van den Brand R., Yew A., Musienko P., Zhong H., Song B., Ao Y., Ichiyama R.M., Lavrov I., Roy R.R., Sofroniew M.V., Edgerton V.R. (2009) Transformation of nonfunctional spinal circuits into functional states after the loss of brain input. *Nat. Neurosci.*, 12: 1333-1342. [Rats with complete spinal cord transections regained the ability to walk on a treadmill after a treatment that combined serotonergic agonists with epidural electrical stimulation]

Dennett D.C. (1992) *Consciousness explained*. 528 pp. Back Bay Books. [This book contains a description of the pioneering demonstration of a brain-machine interface by Grey Walter]

Dobelle W.H., Mladejovsky M.G., Girvin J.P. (1974) Artificial vision for the blind: electrical stimulation of visual cortex offers hope for a functional prosthesis. *Science*, 183: 440-444. [In this pioneering study, electrical stimulation was applied to the occipital cortex through disk electrodes. The stimulation produced phosphenes in two volunteers who had been totally blind for 7 and 28 years, respectively. Moreover, they were able to recognize simple patterns, including letters, produced my stimulation through multiple electrodes]

Ethier C., Oby E.R., Bauman M.J., Miller L.E. (2012) Restoration of grasp following paralysis through brain-controlled stimulation of muscles. *Nature*, 485: 368-371. [The authors developed a cortically driven FES system in monkeys. Forearm muscles were paralyzed by a local anesthetic that blocked the median and ulnar nerves. Intended activity of these muscles was extracted from the firing of approximately 100 neurons in the motor cortex and delivered to the muscles using FES. Aided with this BMI, monkeys were able to perform grasping with their temporarily deinnervated hands.]

Evarts E.V. (1973) Motor cortex reflexes associated with learned movement. *Science*, 179: 501-503. [One of pioneering studies by Edward Evarts on single-unit recordings from cortical neurons in awake, behaving monkeys]

Farwell L.A., Donchin E. (1988) Talking off the top of your head: toward a mental prosthesis utilizing event-related brain potentials. *Electroencephalogr. Clin. Neurophysiol.*, 70: 510-523. [In this pioneering study, a P300 BMI was developed. Subjects communicated through a computer by using the P300 component of the event-related brain potential. A computer screen displayed a matrix of letters and other symbols. The character on which the subject focuses his attention was recognized by repeatedly flashing rows and columns of the matrix]

Farah M.J. (2002) Emerging ethical issues in neuroscience. *Nat. Neurosci.*, 5:1123-1129. [A review of ethical issues raised by progress in many areas of neuroscience, particularly enhancement of normal function and "brain reading"]

Fatourechi M., Bashashati A., Ward R.K., Birch G.E. (2007) EMG and EOG artifacts in brain computer interface systems: A survey. *Clin. Neurophysiol.*, 118: 480-494. [The authors reviewed more than 250 journal and conference papers on EEG-based BMIs and revealed weaknesses related to reporting the methods of handling EMG and EOG artifacts]

Feldman A.G., Ostry D.J., Levin M.F., Gribble P.L., Mitnitski A.B. (1998) Recent tests of the equilibrium-point hypothesis (lambda model). *Motor Control*, 2: 189-205. [A critical review of arguments for and against the equilibrium point hypothesis]

Fetz E.E. (1969) Operant conditioning of cortical unit activity. *Science*, 163: 955-958. [A pioneering report showing that monkeys can learn to volitionally modulate discharge rates of their cortical neurons]

Fernandes R.A., Diniz B., Ribeiro R., Humayun M. (2012) Artificial vision through neuronal stimulation. *Neurosci. Lett.*, 519: 122-128. [A review of significant developments achieved by the visual prosthesis. The authors describe a continued improvement in visual acuity with such devices. Subjects are able to read letters. Their mobility and orientation is improved]

Fitzsimmons N.A., Drake W., Hanson T.L., Lebedev M.A., Nicolelis M.A. (2007) Primate reaching cued by multichannel spatiotemporal cortical microstimulation. *J Neurosci.*, 27: 5593-5602. [This study showed for the first time that information can be delivered to the primary somatosensory cortex using spatiotemporal patterns of intracortical microstimulation]

Fitzsimmons N.A., Lebedev M.A., Peikon I.D., Nicolelis M.A. (2009) Extracting kinematic parameters for monkey bipedal walking from cortical neuronal ensemble activity. *Front. Integr. Neurosci*.3:3. [In this pioneering study researchers extracted kinematic parameters of bipedal locomotion from cortical ensemble activity recorded in monkey brain]

Frank K. (1968) Some approaches to the technical problem of chronic excitation of peripheral nerve. *Ann. Otol. Rhinol. Laryngol.*, 77: 761–771. [One of the earliest discussions on the neural control of prosthetic devices]

Galán F., Nuttin M., Lew E., Ferrez P.W., Vanacker G., Philips J., Millán J.R. (2008) A brain-actuated wheelchair: asynchronous and non-invasive brain-computer interfaces for continuous control of robots. *Clin. Neurophysiol.*, 119: 2159-2169. [This study assessed the feasibility and robustness of an asynchronous and non-invasive EEG-based BMI for continuous mental control of a wheelchair. Subjects were able to rapidly mastered this BMI control because of two key components: a shared control system between the BCI system and the wheelchair; and the selection of stable user-specific EEG features]

Georgopoulos A.P., Lurito J.T., Petrides M., Schwartz A.B., Massey J.T. (1989) Mental rotation of the neuronal population vector. *Science*, 243: 234–236. [A pioneering study describing a population vector approach to the analysis of motor cortical neuronal populations. A rhesus monkey was trained to perform arm reaching movements in a direction that was perpendicular to a visual target location. The population vector constructed from the activities of individual neurons rotated gradually from the direction of the target to the direction of the movement]

Guertin P.A. (2009) The mammalian central pattern generator for locomotion. *Brain Res. Rev.*, 62: 45-56. [A comprehensive review of central pattern generators]

Hanson T.L., Fuller A.M., Lebedev M.A., Turner D.A., Nicolelis M.A.L. (2012) Subcortical neuronal ensembles: an analysis of motor task association, tremor, oscillations, and synchrony in human patients. *J. Neurosci.*, 32: 8620-8632. [This paper explores the possibility of gaining a BMI control signal from subcortical areas in human subjects]

Hatsopoulos N.G., Donoghue J.P. (2009) The science of neural interface systems. *Annu. Rev. Neurosci.*, 32: 249-266. [A review of brain-machine interfaces based on the recordings of single-unit action potentials from the brain]

Haykin S. (2001) *Adaptive Filter Theory (4th ed.)*, 936 pp. Upper Saddle River, New Jersey, Prentice Hall. [A comprehensive book on adaptive filters. These filters are used as decoding algorithms for brain-machine interfaces]

Head H., Holmes, G. (1911) Sensory disturbances from cerebral lesions. *Brain*, 34: 102-254. [A pioneering study that proposed a concept of body schema -- a representation of the body, built, maintained and updated by the brain. The schema is used to interpret sensory signals from peripheral sensory receptors and to give rise to perceptions correctly mapped to the body configuration. Body schema is also very important for programming and executing movements]

Hochberg L.R., Serruya M.D., Friehs G.M., Mukand J.A., Saleh M., Caplan A.H., Branner A., Chen D., Penn R.D., Donoghue J.P. (2006) Neuronal ensemble control of prosthetic devices by a human with

tetraplegia. *Nature*, 442: 164-171. [This paper reports initial results for a tetraplegic human implanted with a multielectrode array in the motor cortex. The patient was able to control a computer cursor, navigate through e-mail and open and close a hand prosthesis]

Hochberg L.R., Bacher D., Jarosiewicz B., Masse N.Y., Simeral J.D., Vogel J., Haddadin S., Liu J., Cash S.S., van der Smagt P., Donoghue J.P. (2012) Reach and grasp by people with tetraplegia using a neurally controlled robotic arm. *Nature*, 485: 372-375. [This paper demonstrates an invasive BMI that enabled people with long-standing tetraplegia to perform reach and grasp movements with a robotic arm]

Hubel D.H., Wiesel T.N. (2005). *Brain and visual perception: the story of a 25-year collaboration*. 744 pp. Oxford University Press. [A story of pioneering studies on the visual cortex by two Nobel Prize winners]

Humphrey D.R., Schmidt E.M., Thompson W.D. (1970) Predicting measures of motor performance from multiple cortical spike trains. *Science*, 170: 758-762. [The first demonstrations that movement parameters can be extracted from the activity of small populations of motor cortical neurons]

Iriki A., Tanaka M., Iwamura Y. (1996) Coding of modified body schema during tool use by macaque postcentral neurones. *Neuroreport*, 7: 2325-2330. [This study demonstrated brain plasticity associated with tool use. Macaque monkeys were trained to retrieve distant objects using a rake. During tool use, visual receptive fields of posterior parietal cortex neurons were altered to cover the expanded accessible space]

Jones L.A. (2011) Tactile communication systems optimizing the display of information. *Prog. Brain Res.*, 192:113-128. [This review discusses tactile communication systems based on vibrotactile signals. Such systems are used as sensory substitution devices for people with visual, auditory, or vestibular impairments]

Kalaska JF, Scott SH, Cisek P, Sergio LE (1997) Cortical control of reaching movements. *Curr. Opin. Neurobiol.*, 7: 849-859. [A review of neurophysiological literature on the involvement of different cortical areas in the process of transformation of sensory stimuli into movements]

Kawato M. (1999) Internal models for motor control and trajectory planning. *Curr. Opin. Neurobiol.*, 9: 718–727. [A review of kinematic and dynamic internal models in neural circuits that subserve motor control]

Kennedy P.R., Bakay RA. (1998) Restoration of neural output from a paralyzed patient by a direct brain connection. *Neuroreport*, 9: 1707-1711. [A BMI study in a 'locked-in' patient with ALS who was implanted with an electrode that induces growth of myelinated fibers into its recording tip. The patient was able to communicate in an on/off fashion using this BMI]

Lebedev M.A., Carmena J.M., O'Doherty J.E., Zacksenhouse M., Henriquez C.S., Principe J.C., Nicolelis M.A. (2005) Cortical ensemble adaptation to represent velocity of an artificial actuator controlled by a brain-machine interface. *J. Neurosci.*, 25: 4681-4693. [This paper examines plastic changes in cortical ensemble activity associated with direct brain control of a robotic arm. The authors suggests that cortical networks assimilated the robot as a new extension of the body]

Lebedev M.A., Nicolelis M.A. (2006) Brain-machine interfaces: past, present and future. *Trends Neurosci.*, 29: 536-546. [This review covers invasive and noninvasive brain-machine interfaces and main problems in their development]

Lebedev M.A., Nicolelis M.A. (2011) Toward a whole-body neuroprosthetic. *Prog. Brain Res.*, 194: 47-60. [A review of steps that need to be taken to build a whole-body BMI that will incorporate very large-scale brain recordings, advanced decoding algorithms, artificial sensory feedback based on electrical stimulation of somatosensory areas, virtual environment representations, and a whole-body exoskeleton]

Li Z., O'Doherty J.E., Hanson T.L., Lebedev M.A., Henriquez C.S., Nicolelis MA (2009) Unscented Kalman filter for brain-machine interfaces. *PLoS One*, 4: e6243. [This paper introduces a new algorithm, an n-th order unscented Kalman filter, which enhances the standard Kalman filter with two innovations: non-linear model of neural tuning and augmentation of the movement state variables with a time history of neuronal discharges. The new filter was implemented in BMI experiments where it outperformed the standard Kalman filter and the Wiener filter]

Lin C.T., Chang C.J., Lin B.S., Hung S.H., Chao C.F., Wang I.J. (2010). A Real-Time Wireless Brain-Computer Interface System for Drowsiness Detection. *Biomedical Circuits and Systems, IEEE Transactions on*, 4(4), 214-222. [A report of a real-time wireless EEG-based BMI for drowsiness detection. This module is small enough to be embedded into a headband as a wearable EEG device that can be used in practical driving]

Lilly J.C. (1956) Distribution of 'motor' functions in the cerebral cortex in the conscious, intact monkey. *Science*, 124: 937. [The first report of large number of electrodes implanted in monkey cortex]

McFarland D.J., Krusienski D.J., Wolpaw J.R. (2006) Brain-computer interface signal processing at the Wadsworth Center: mu and sensorimotor beta rhythms. *Prog. Brain Res.*, 159: 411-419. [A review of signal decoding in EEG-based brain-machine interfaces]

Mellinger J., Schalk G., Braun C., Preissl H., Rosenstiel W., Birbaumer N., Kübler A. (2007). An MEGbased brain–computer interface (BCI). *Neuroimage*, 36: 581-593. [This study investigated the utility of a magnetoencephalograhy (MEG)-based BMI. MEG recordings detected voluntary modulations of sensorimotor μ and β rhythms. Subjects achieved significant BMI control within 32 min of training]

Millan J.R., Renkens F., Mouriño J., Gerstner W. (2004) Noninvasive brain-actuated control of a mobile robot by human EEG. *IEEE Trans. on Biomed. Eng.*, 51: 1026-1033.

Moritz C.T., Perlmutter S.I., Fetz E.E. Direct control of paralysed muscles by cortical neurons. *Nature*, 456: 639-642. [This paper shows that monkeys can directly control electrical stimulation of their temporarily paralyzed muscles by individual cortical neurons]

Mountcastle V.B. (2005) *The sensory hand: neural mechanisms of somatic sensation.* 640 pp. Cambridge, MA, Harvard University Press. [This book by Vernon Mountcastle reviews his views on brain mechanisms of somatosensory sensation]

Muller-Putz G.R., Pfurtscheller G. (2008) Control of an electrical prosthesis with an SSVEP-based BCI. *IEEE Trans. Biomed. Eng.*, 55: 361–364. [Subjects used a BCI to control a two-axes electrical hand prosthesis with the flickering lights mounted on its surface]

Nicolas-Alonso L.F., Gomez-Gil J. (2012) Brain computer interfaces, a review. *Sensors (Basel)*, 12: 1211-1279. [A review of invasive nonivasive BMIs that discusses advantages and drawbacks of different systems, methods for acquisition electrical, magnetic or metabolic activity and decoding approaches]

Nicolelis M.A., Lebedev M.A. (2009) Principles of neural ensemble physiology underlying the operation of brain-machine interfaces. *Nat. Rev. Neurosci.*, 10, 530-540. [This article reviews brain-machine interfaces and principles of distributed neural processing that they reveal]

Nicolelis M.A. (2011) *Beyond Boundaries: The New Neuroscience of Connecting Brains with Machines – and How It Will Change Our Lives*, 354 pp. New York, New York, Times Books. [This book presents a very comprehensive discussion of the history and future directions of brain-machine interfaces]

O'Doherty J.E., Lebedev M.A., Ifft P.J., Zhuang K.Z., Shokur S., Bleuler H., Nicolelis M.A. (2011) Active tactile exploration using a brain-machine-brain interface. *Nature*, 479: 228-231. [This study pioneered a brain-machine-brain interface, an interface that both reads motor information from the brain and delivers artificial tactile feedback to the brain using intracortical microstimulation]

Pfurtscheller G., Müller G.R., Pfurtscheller J., Gerner H.J., Rupp R. (2003) 'Thought'--control of functional electrical stimulation to restore hand grasp in a patient with tetraplegia. *Neurosci. Lett.*, 351:33-36. [This study demonstrated for the first time that hand grasp function in a tetraplegic patient can be restored using functional electrical stimulation driven by EEG recordings]

Pohlmeyer E.A., Oby E.R., Perreault E.J., Solla S.A., Kilgore K.L., Kirsch R.F., Miller L.E. (2009) Toward the restoration of hand use to a paralyzed monkey: brain-controlled functional electrical stimulation of forearm muscles. *PLoS One*, 4: e5924. [The authors developed a system that used neural signals recorded in the motor cortex of rhesus monkeys to control FES applied to forearm muscles. Two monkeys used this FES-based BMI to contract muscles despite temporary limb paralysis]

Presacco A., Forrester L.W., Contreras-Vidal J.L. (2012) Decoding intra-limb and inter-limb kinematics during treadmill walking from scalp electroencephalographic (EEG) signals. *IEEE Trans. Neural Syst. Rehabil. Eng.*, 20: 212-219. [In this study, gait kinematics was extracted from EEG signals sampled using as few as 12 electrodes]

Quiroga R.Q., Reddy L., Kreiman G., Koch C., Fried I. (2005) Invariant visual representation by single neurons in the human brain. *Nature*, 435: 1102-1107. [The authors report a subset of neurons in the human medial temporal lobe that are selectively activated by pictures of individuals, for example "Jennifer Aniston cells"]

Romo R., Hernández A., Zainos A., Brody C.D., Lemus L. (2000) Sensing without touching: psychophysical performance based on cortical microstimulation. *Neuron*, 26: 273-278. [This study showed that microstimulation of primary somatosensory cortex can be used to elicit discriminable percepts that resemble those evoked by vibrotactile stimulation]

Sampaio E., Maris S., Bach-y-Rita P. (2001) Brain plasticity: 'visual' acuity of blind persons via the tongue. *Brain Res.*, 908: 204-207. [This article reports a sensory substitution device that restores vision by electrical stimulation of the tongue]

Schmidt E.M. (1980) Single neuron recording from motor cortex as a possible source of signals for control of external devices. *Ann. Biomed. Eng.* 8: 339-349. [This paper reports chronic recordings of monkey cortical neurons for as long as 37 months. Monkeys learned to move a cursor on a LED display by modulating their neurons]

Schott G.D. (1993). Penfield's homunculus: a note on cerebral cartography. *Journal of Neurology, Neurosurgery, and Psychiatry*, 56: 329-333. [A review of how the idea of cortical homunculus developed after it was first proposed by Penfield, and whether or not it helped to understand cortical representation of the body]

Schwartz A.B., Cui X.T., Weber D.J., Moran D.W. (2006) Brain-controlled interfaces: movement restoration with neural prosthetics. *Neuron*, 52: 205-220. [A review of tissue-electrode interface, electrode design, and extraction algorithms in brain-machine interfaces]

Sellers E.W., Vaughan T.M., Wolpaw J.R. (2010) A brain-computer interface for long-term independent home use. *Amyotroph. Lateral Scler.*, 11: 449–455. [A report of a reliable noninvasive BMI suitable for long term home use by severely disabled people]

Shannon R.V. (2012) Advances in auditory prostheses. *Curr. Opin. Neurol.*, 25: 61-66. [A review of auditory prostheses, both at the level of the sensory nerve and the brainstem]

Sherrington C.S. (1906) *The integrative action of the nervous system*. 411 pp. Liverpool. New York, Charles Scribner's Sons. [In this classical book, Sherrington describes his theory of neural control of posture and movements]

Sitaram R., Caria A., Birbaumer N. (2009) Hemodynamic brain-computer interfaces for communication and rehabilitation. *Neural Netw.*, 22: 1320-1328. [A review of BMIs that utilize the measurement of the blood oxygen level in the brain: near-infrared spectroscopy (NIRS) and functional magnetic resonance imaging (fMRI) based BMIs]

Sussillo D., Nuyujukian P., Fan J.M., Kao J.C., Stavisky S.D., Ryu S., Shenoy K. (2012) A recurrent neural network for closed-loop intracortical brain-machine interface decoders. *J Neural Eng.*, 9: 026027. [This paper introduces a new BMI decoding method that outperforms previous decoders]

Tangermann M., Krauledat M., Grzeska K., Sagebaum M., Blankertz B., Vidaurre C., Müller K.R. (2009) Playing pinball with non-invasive BCI. *Advances in Neural Information Processing Systems*, 21: 1641-1648. [Subjects used an EEG-based BMI to play on a pinball machine. Fast and well-timed control was achieved using machine learning methods for mental state decoding]

Tavella M., Leeb R., Rupp R., Millán J.d.R. (2010). Towards natural non-invasive hand neuroprostheses for daily living. *Proceedings of the 32nd Annual International Conference of the IEEE Engineering in Medicine and Biology Society, Buenos Aires*. [In this paper subjects operated a noninvasive BMI that controls a functional electrical stimulator of forearm muscles. They were able to perform such sophisticated tasks as handwriting]

Taylor D.M., Tillery S.I., Schwartz A.B. (2002) Direct cortical control of 3D neuroprosthetic devices. Science 296: 1829-1832. [A study of 3-dimensional control of a virtual-environment cursor by motor cortical populations recorded in awake monkeys. A coadaptive algorithm was implemented to attune the BMI performance when the monkeys operated the BMI without moving their arms]

Velliste M., Perel S., Spalding M.C., Whitford A.S., Schwartz A.B. (2008) Cortical control of a prosthetic arm for self-feeding. *Nature*, 453: 1098-1101. [In this study, monkeys were implanted with recording arrays in the motor cortex. A BMI was implemented that enabled the animals to feed themselves with a robotic arm.]

Vialatte F.B., Maurice M., Dauwels J., Cichocki A. (2010) Steady-state visually evoked potentials: focus on essential paradigms and future perspectives. *Prog. Neurobiol.*, 90: 418-438. [A review of applications steady-state visually evoked potentials (SSVEPs) in Neuroscience and BMI research]

Vlek R.J., Steines D., Szibbo D., Kübler A., Schneider M.J., Haselager P., Nijboer F. (2012) Ethical issues in brain-computer interface research, development, and dissemination. *J. Neurol. Phys. Ther.*, 36: 94-99. [A review of ethical issues emerging from the steadily growing field of BMIs. Four examples of clinical scenarios are discussed.]

Wessberg J., Stambaugh C.R., Kralik J.D., Beck P.D., Laubach M., Chapin J.K., Kim J., Biggs S.J., Srinivasan M.A., Nicolelis M.A. (2000) Real-time prediction of hand trajectory by ensembles of cortical neurons in primates. *Nature*, 408: 361-365. [In this pioneering study, real-time control of a robotic arm was produced by simultaneous activity of large populations of monkey cortical neurons processed by linear and nonlinear algorithms]

Wilson B.S., Dorman M.F. (2008) Cochlear implants: a remarkable past and a brilliant future. *Hear. Res.*, 242: 3-21. [This review article provides a brief history of cochlear implants, describes the current state of implant technology and discusses limitations of current devices]

Wise S.P. (1985) The primate premotor cortex: past, present, and preparatory. *Annu. Rev. Neurosci*, 8: 1-19. [A review on the representation of motor preparatory signals in the premotor cortex]

Wolpaw J.R., Birbaumer N., McFarland D.J., Pfurtscheller G., Vaughan T.M. (2002) Brain-computer interfaces for communication and control. *Clin. Neurophysiol.*, 113: 767-791. [A review of BMIs based on a conversion of EEGs into communication channel for sending messages and commands to the external world. The authors conclude that such interfaces could eventually provide new communication and control options for people with motor disabilities and might also give those without disabilities supplementary channels for special circumstances]

Wolpaw J.R., McFarland D.J. (2004) Control of a two-dimensional movement signal by a noninvasive brain-computer interface in humans. *Proc. Natl. Acad. Sci. USA*, 101: 17849-17854. [A demonstration of EEG-based brain-machine interface that continuously controlled cursor movements on a computer screen. Human subjects were able to perform center-out reaching movements with good accuracy]

Zacksenhouse M., Lebedev M.A., Carmena J.M., O'Doherty J.E., Henriquez C., Nicolelis M.A. (2007) Cortical modulations increase in early sessions with brain-machine interface. *PLoS One*, 2: e619. [This study examined neuronal plasticity during the operation of brain-machine interfaces. Monkey cortical neurons transiently increased their modulations during learning to control a brain-machine interface that enacted arm reaching movements]

Zhang F., Aravanis A.M., Adamantidis A., de Lecea L., Deisseroth K. (2007) Circuit-breakers: optical technologies for probing neural signals and systems. *Nat. Rev. Neurosci.*, 8: 577-581. [This article reviews optogenetics: a newly developed neuroengineering tool based on two microbial opsins, channelrhodopsin-2 (ChR2) and halorhodopsin (NpHR). Optogenetics enables cell-type-specific neuromodulation of neural circuitry]

Biographical Sketch

Mikhail Lebedev received his undergraduate degree in Physics from Moscow Institute of Physics and Technology and his PhD degree in Neurobiology in University of Tennessee, Memphis. He conducted research on motor control and neurophysiology at the Institute for Information Transmission Problems, Moscow, University of Tennessee, Memphis, La Scuola Internazionale Superiore di Studi Avanzati, Trieste and National Institute of Mental Health, Bethesda. He is currently a Senior Research Scientist at Duke University Center for Neuroengineering. Research interests include system neurosciences, primate motor control and cognition and brain-machine interfaces.