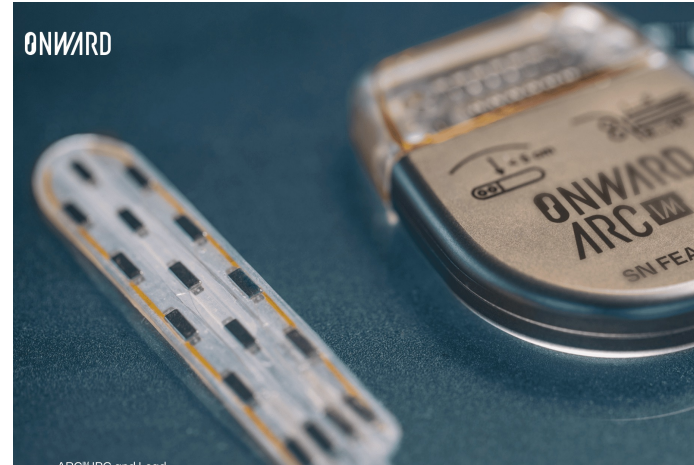
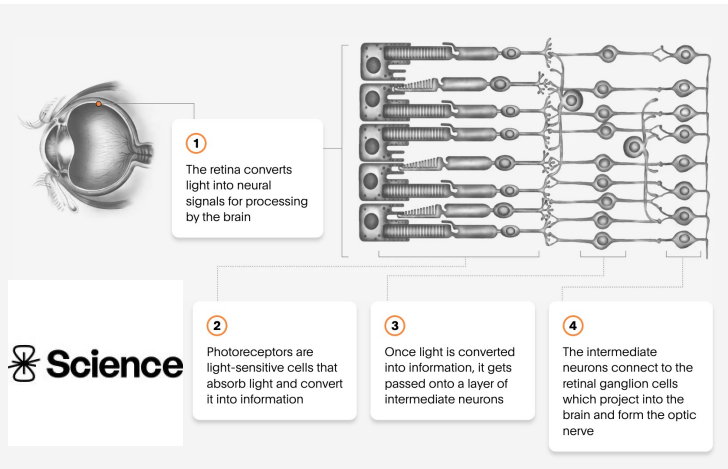


NX-435

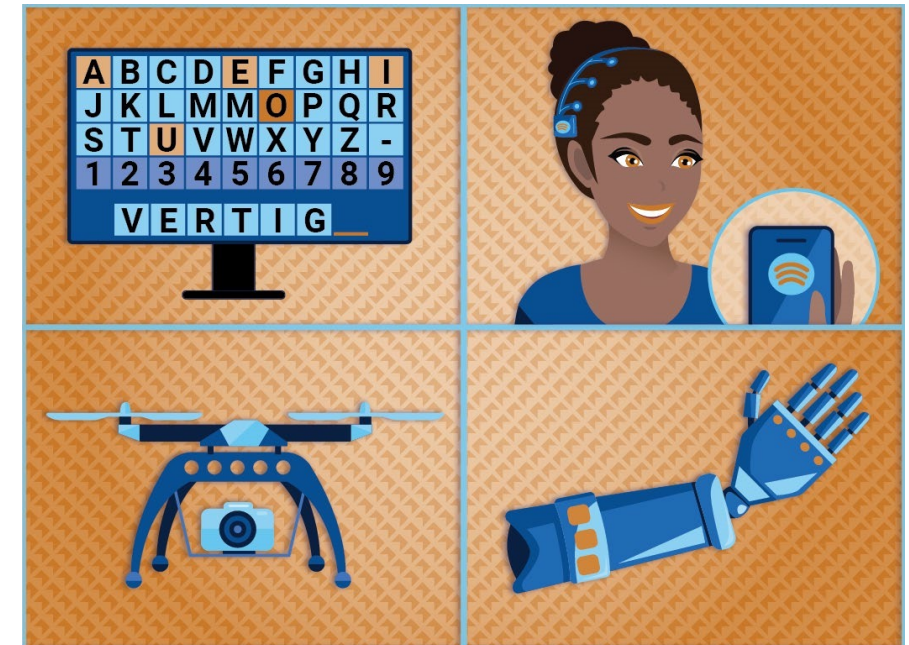
Brain machine interfaces for systems neuroscience

Prof. Mackenzie Mathis, PhD &
Spencer Bowles, PhD

Brain computer interfaces in industry



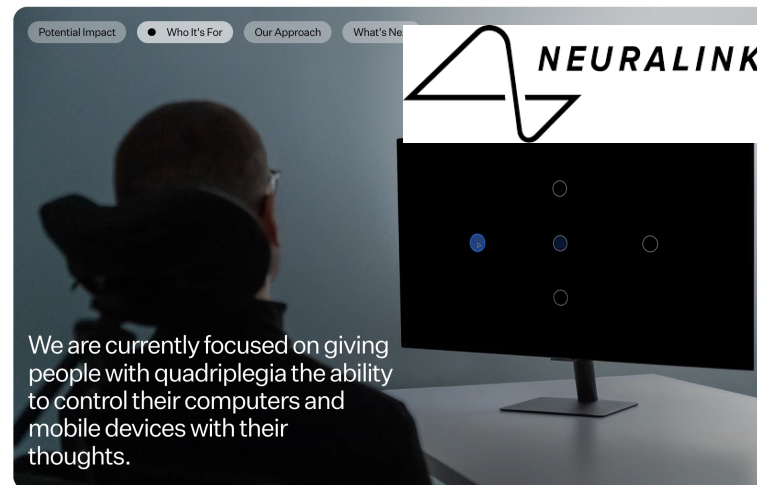
Simply put, a brain-computer interface (BCI) enables a person to control an external device using brain signals



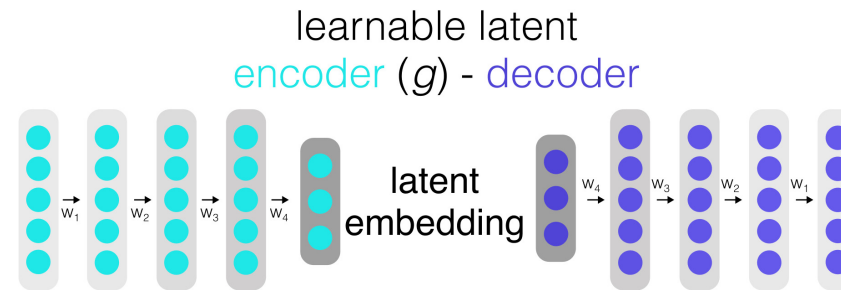
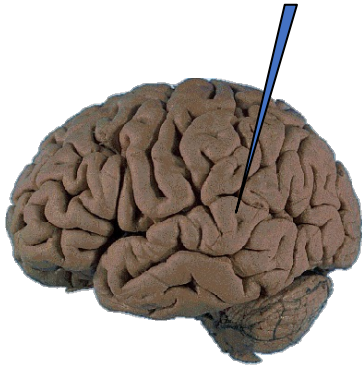
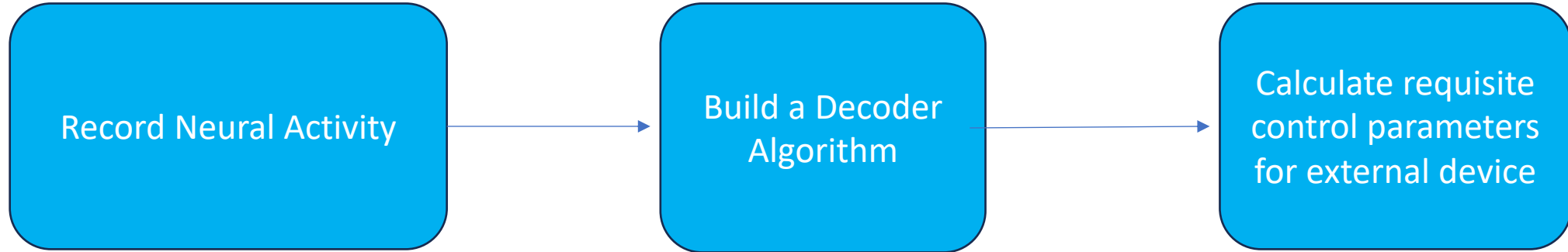
Source: GAO analysis (data). emojoez/svitlana/titaporn/stock.adobe.com (images). | GAO-22-106118

<https://www.gao.gov/products/gao-22-106118>

tech at Meta



Simple overview



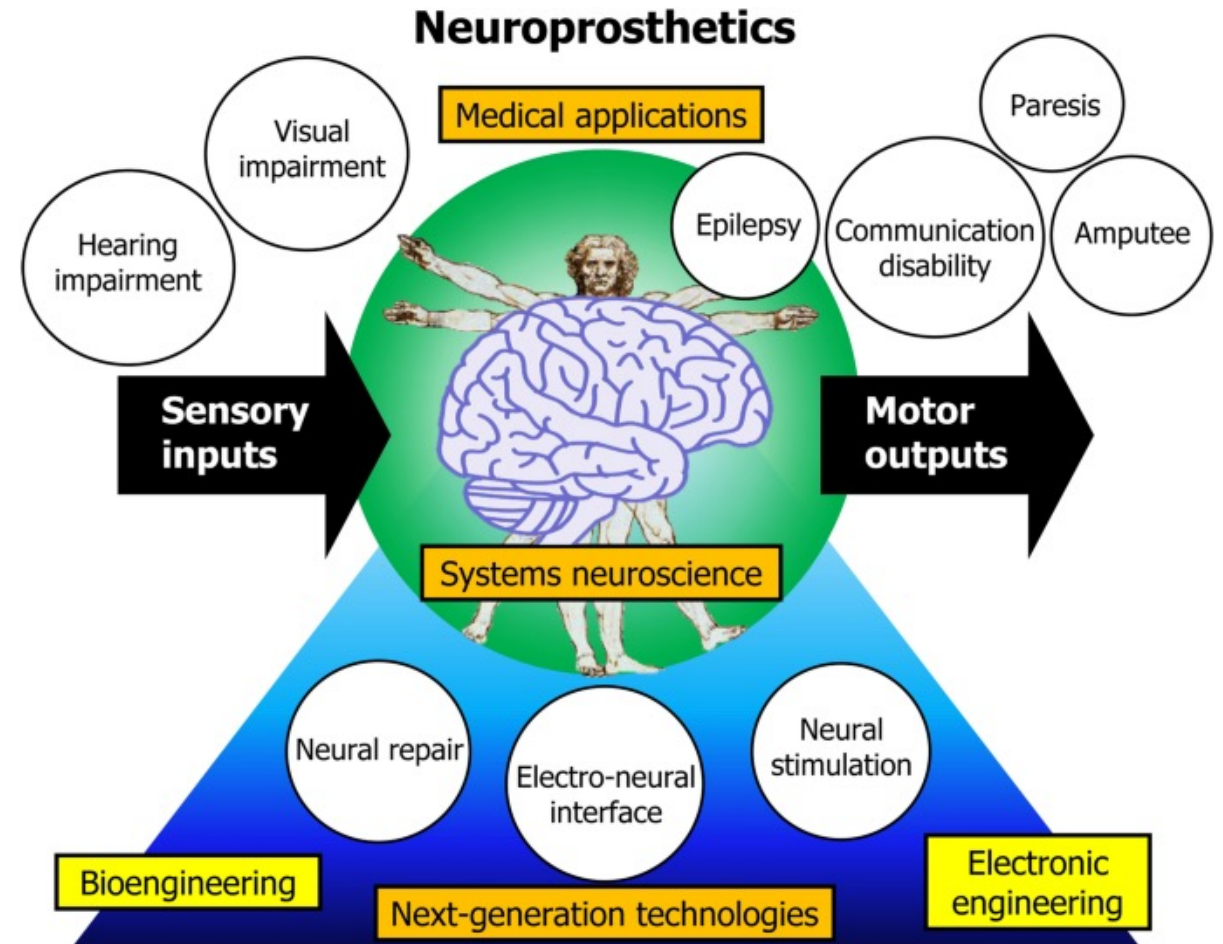
How systems neuroscience is enabling advancements in neuroprosthetics & BCIs

Neuroprosthetics in systems neuroscience and medicine

Our accumulating knowledge in systems neuroscience combined with the development of innovative technologies may enable brain restoration for patients with nervous system disorders. This Collection provides a platform for interdisciplinary research in neuroprosthetics. It will gather studies investigating medical applications of systems neuroscience, informatics, and engineering in the development of neural prostheses. Submissions with a clinical focus on nervous system diseases and brain repair in either humans or animals are also included.

<https://www.nature.com/collections/hjcgjcach>

- Which brain areas to record from
- Need to understand neural subtypes
- How to give appropriate sensory feedback
- How do we enable adaptation and learning



Systems neuroscience and neuroprosthetics

Article

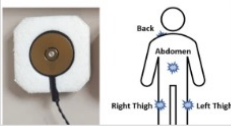
Open Access

17 Aug 2022

Scientific Reports

Identifying potential training factors in a vibrotactile P300-BCI

M. Eidel & A. Kübler



Article

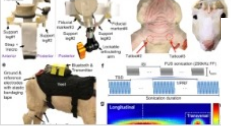
Open Access

29 Sep 2021

Scientific Reports

Transcranial focused ultrasound modulates cortical and thalamic motor activity in awake sheep

Hyun-Chul Kim, Wonhye Lee ... Seung-Schik Yoo



Article

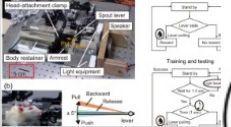
Open Access

23 Sep 2021

Scientific Reports

Prepulse inhibition predicts subjective hearing in rats

Naoki Wake, Kotaro Ishizu ... Hirokazu Takahashi



Article

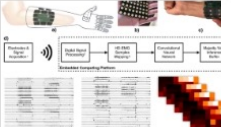
Open Access

28 May 2021

Scientific Reports

Intuitive real-time control strategy for high-density myoelectric hand prosthesis using deep and transfer learning

Simon Tam, Mounir Boukadoum ... Benoit Gosselin



Article

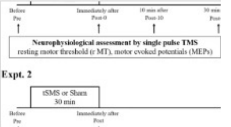
Open Access

8 Mar 2021

Scientific Reports

Transcranial static magnetic stimulation over the motor cortex can facilitate the contralateral cortical excitability in human

Yasuyuki Takamatsu, Satoko Koganemaru ... Tatsuya Mima



Article

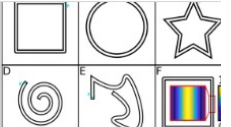
Open Access

25 Feb 2021

Scientific Reports

Quantifying the alignment error and the effect of incomplete somatosensory feedback on motor performance in a virtual brain-computer-interface setup

Robin Lienkämper, Susanne Dyck ... Christian Klaes



Article

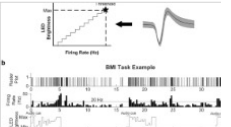
Open Access

17 Nov 2020

Scientific Reports

Operant conditioning of motor cortex neurons reveals neuron-subtype-specific responses in a brain-machine interface task

Martha Gabriela Garcia-Garcia, Cesar Marquez-Chin & Milos R. Popovic



Article

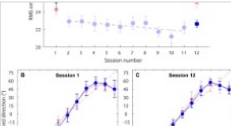
Open Access

25 Aug 2020

Scientific Reports

Haptic sound-localisation for use in cochlear implant and hearing-aid users

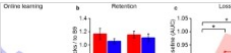
Mark D. Fletcher & Jana Zgheib

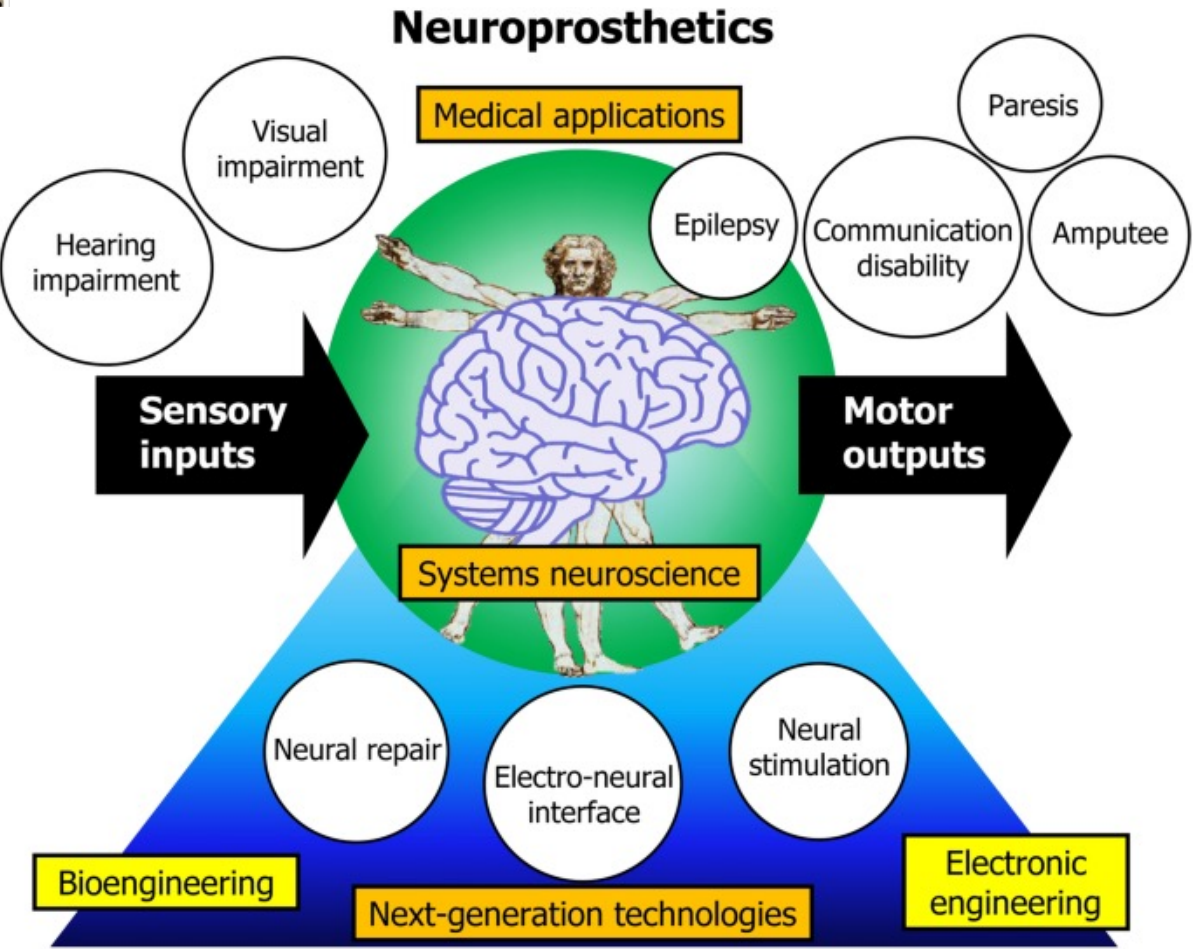


Article

Open Access

Cerebellar transcranial alternating current stimulation in the gamma range applied during the acquisition of a novel motor skill



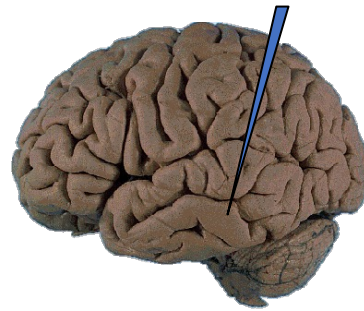


Brain computer interfaces: a primer in decoding

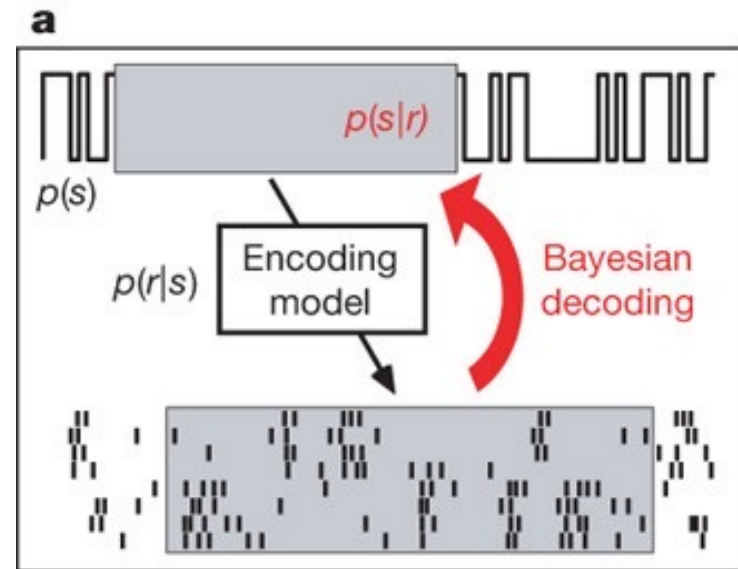
Stimulus



Spikes

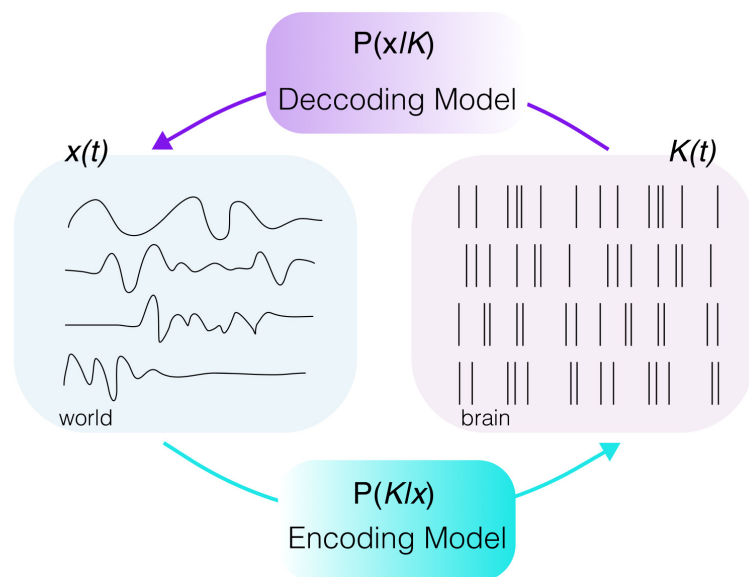


- Our brain needs to determine what is going on in the real world from patterns of spikes.



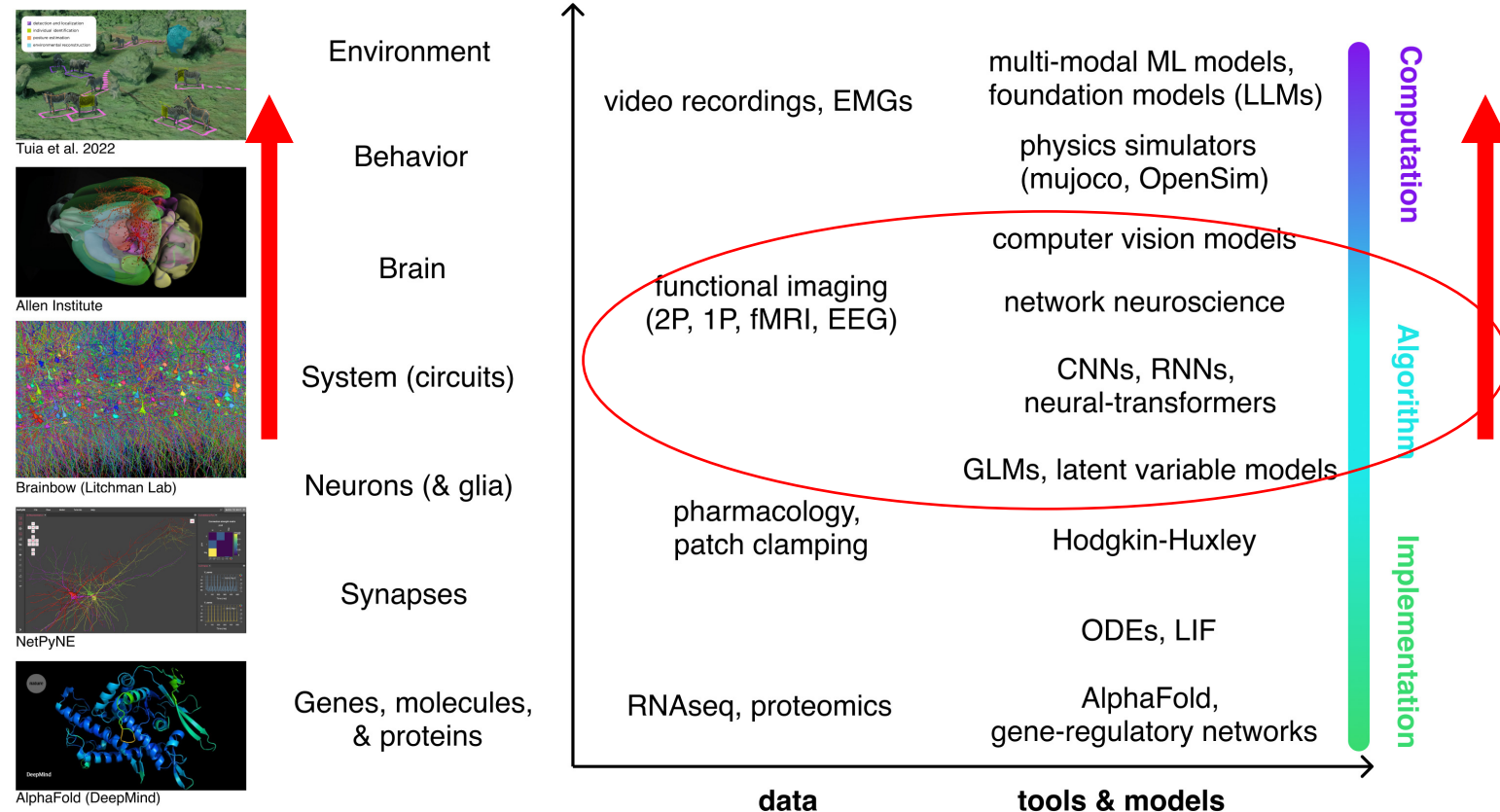
Brain computer interfaces: a primer in decoding

A encoding and decoding models



An encoder represents the neural response of population $K(t)$ to stimulus $x(t)$ via $P(K|x)$, and a decoder aims to recover $x(t)$ given the neural activity $K(t)$ via $P(x|K)$.

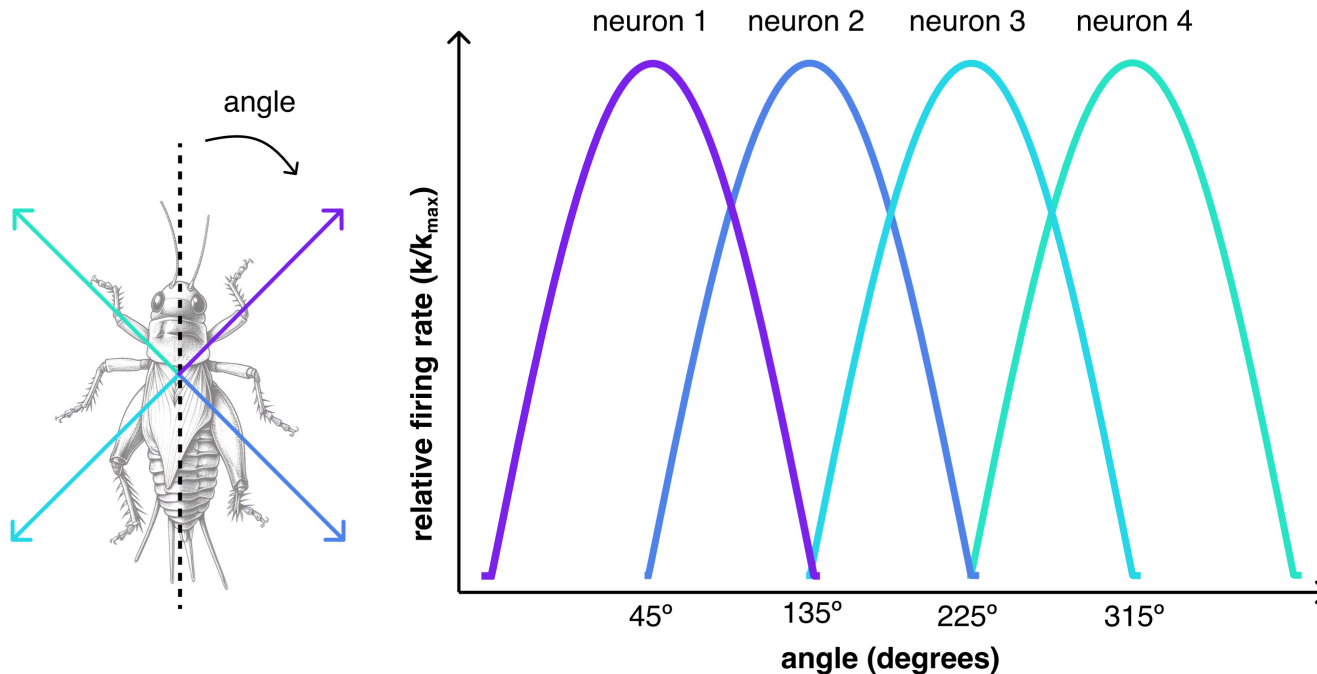
B levels of analysis



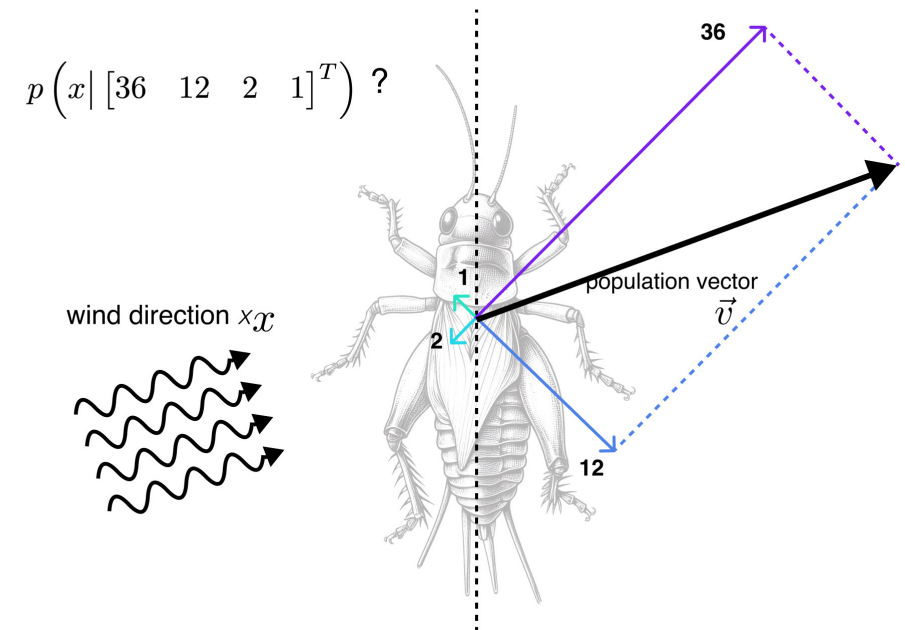
Systems neuroscience spans scales of descriptions and decoding algorithms can target any individual level, and even span across scales. Here, we outline example scales (from genes to environment), the types of data we can collect (from genetic sequencing to whole-animal video analysis), and the classes of models the field has developed. On the far right is our mapping of scales, example data, and example tools to levels of understanding (akin to Marr's 'three levels' (14), with a systems perspective)

Brain computer interfaces: a primer in decoding

A encoding model of wind direction



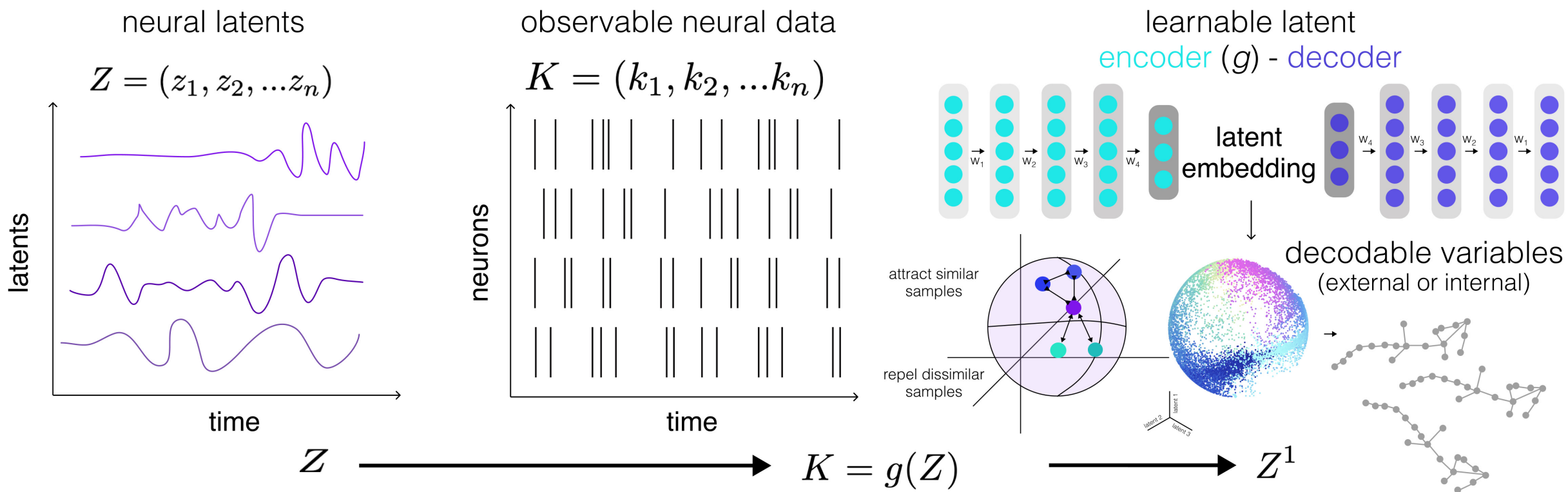
B decoding with population vector



A: The cercal system of the cricket has four interneurons that represent the wind direction. The preferred wind directions of the neurons are pointing in four cardinal directions and can be represented by orthogonal vectors (on the left). Each neuron responds with a firing rate approximated by a half-wave rectified cosine function. The maximum firing rate is elicited when the wind is blowing in the preferred direction.

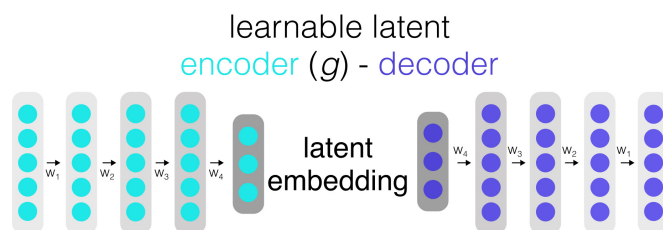
B: The wind direction x can be decoded as the direction of the population vector \hat{x} . This vector is the sum of the four preferred orientations scaled by their firing rate. An example is shown for neurons responding with activity $[36, 12, 2, 1]^T$. Note how the population vector closely matches the wind direction

Brain computer interfaces: a primer in decoding



Learnable latent variable models. A: On the path to building more causal models are new frameworks, such as CEBRA (27), that allow for learning a mapping from the observable data K to the latent dynamics Z . Here, the aim is to use identifiable models with contrastive learning $g(Z)$ (the encoder), then invert this model or use another decoder framework to probe the relationship between the estimated latents, Z^1 , and a variable such as an externally observable state (behavior), internal, or sensory (i.e., recover some stimulus space (x)).

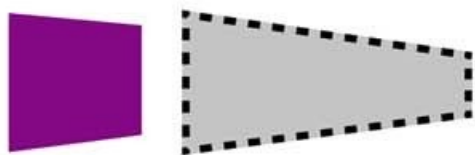
Brain computer interfaces: a primer in decoding



adaptation to a new session

pretrained

- Input embedding is finetuned
- Feature extractor is fixed (pretrained)

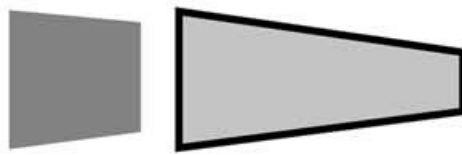


- Input embedding is finetuned
- Feature extractor is finetuned

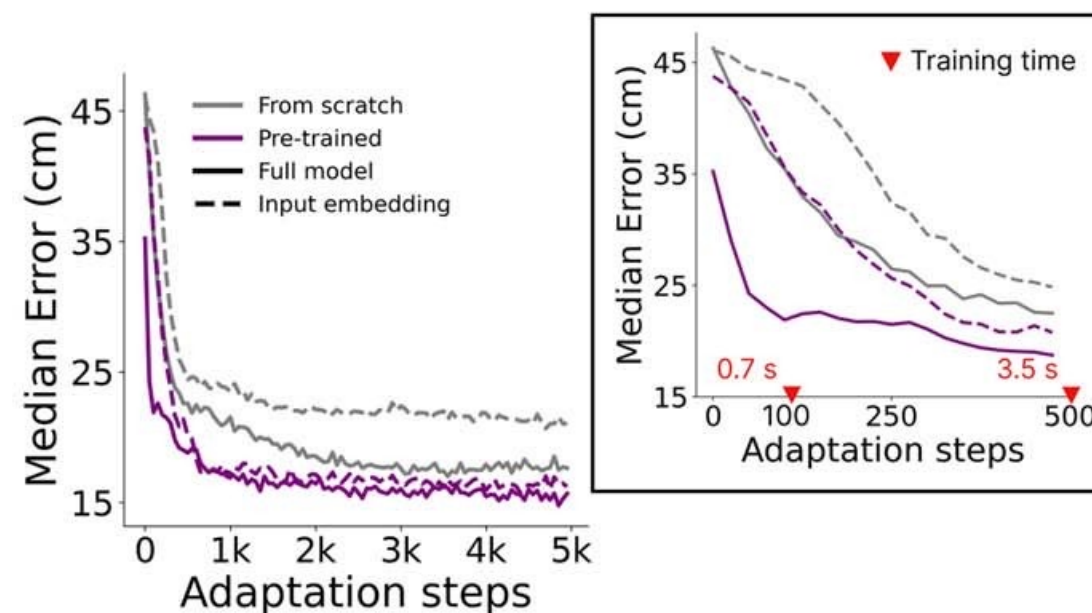
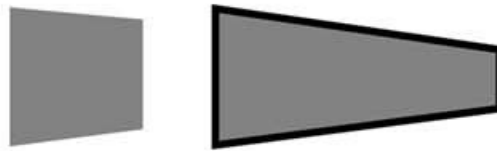


scratch

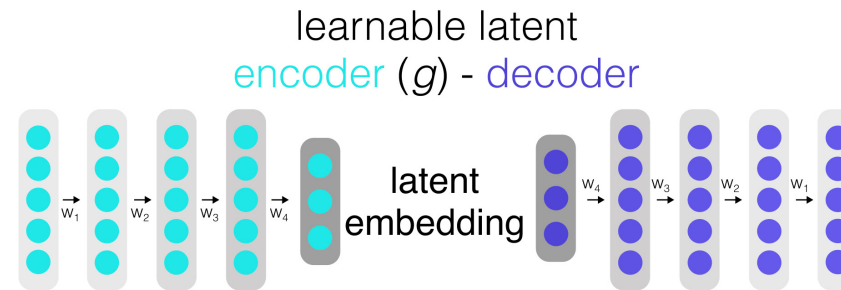
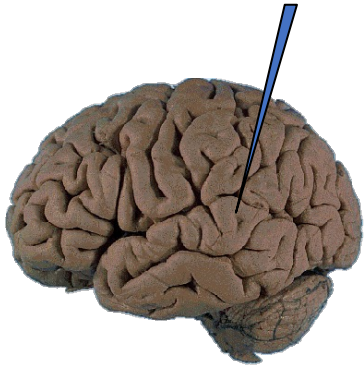
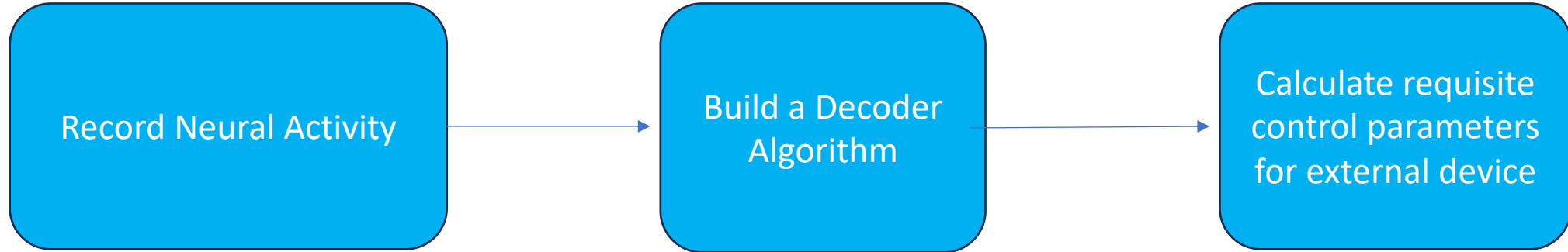
- Input embedding is finetuned
- Feature extractor is from scratch



- Full model is finetuned
- Feature extractor is from scratch

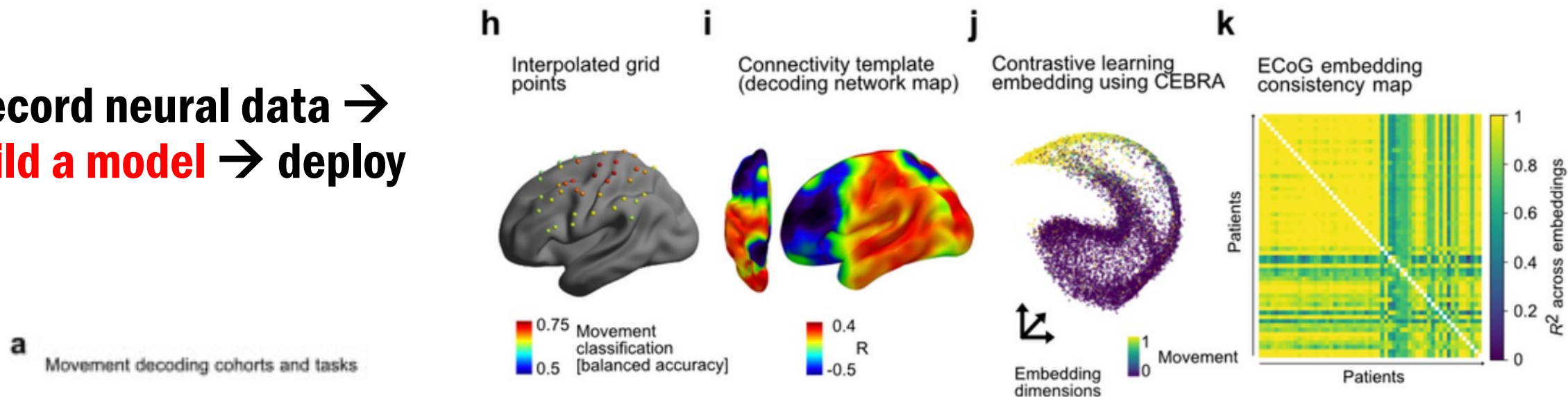


Simple overview



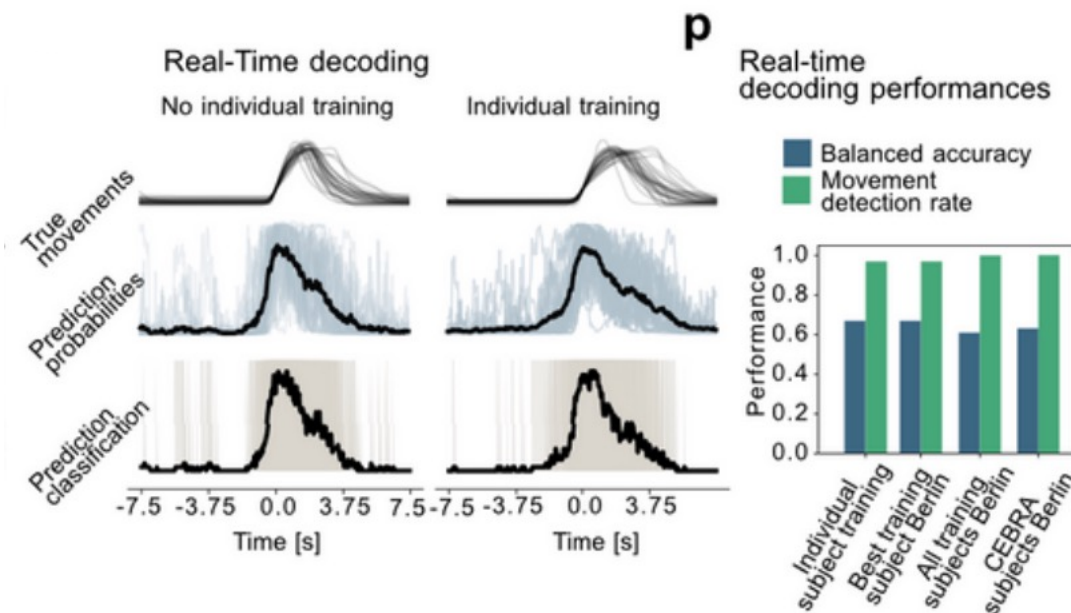
Record neural data →
build a model → deploy

Methods for decoding without patient individual training



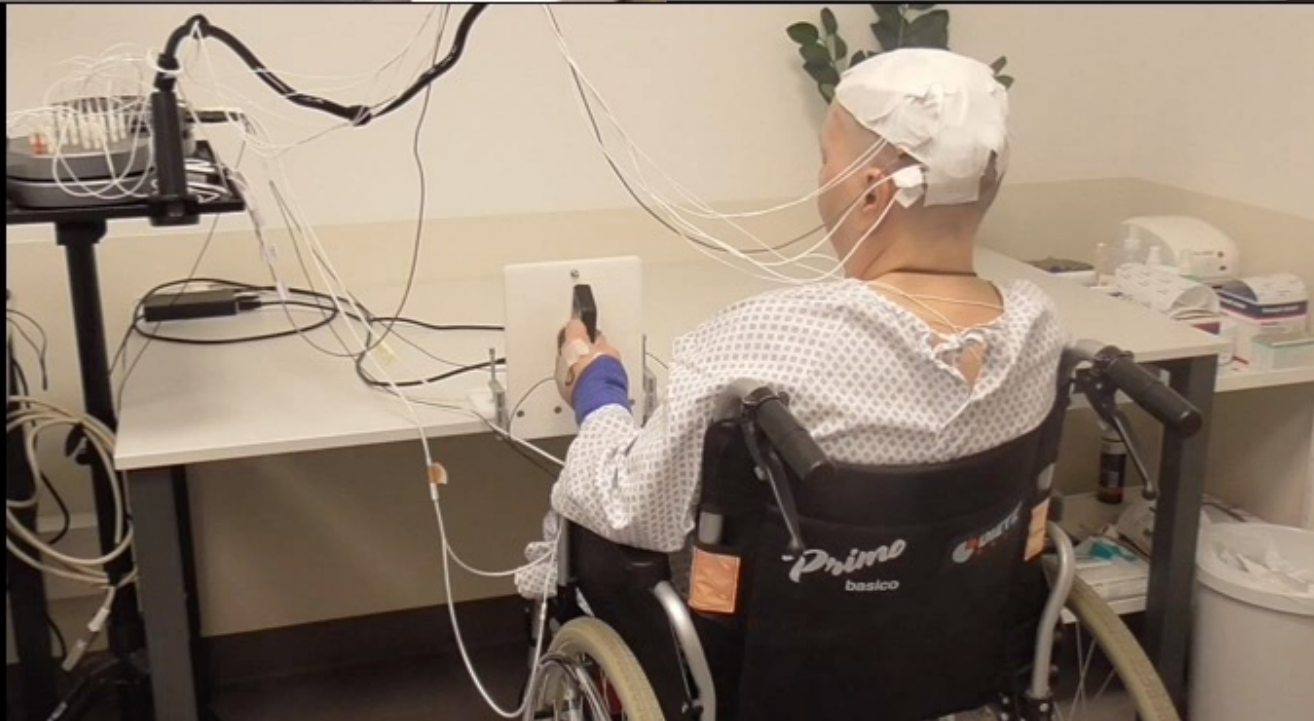
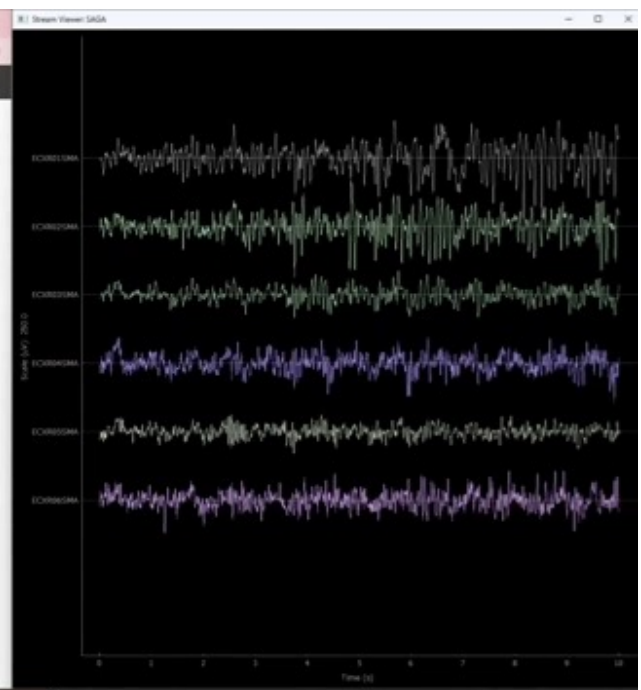
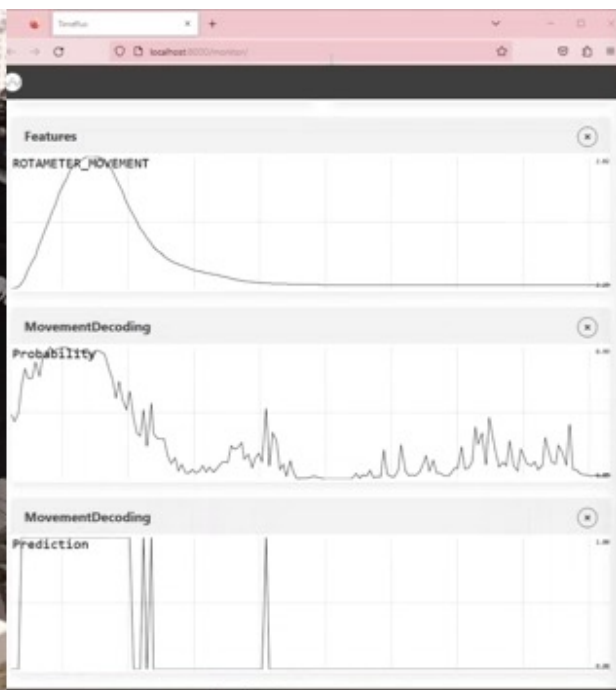
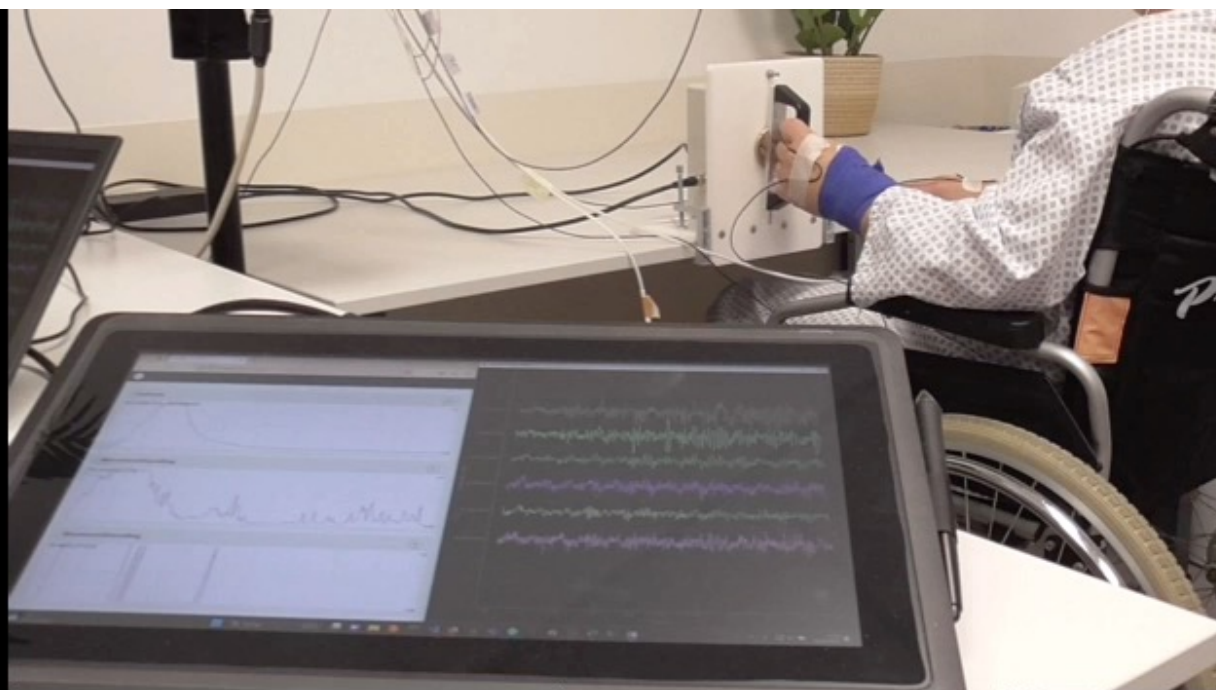
Center	Berlin	Beijing	Pittsburgh	Washington
Subjects	12 PD	10 PD	16 PD	18 Epilepsy
Task	Rotation	Button press	Grip force	Clench & release

Plug & play decoding on a newly recruited subject



Merk et al. 2023

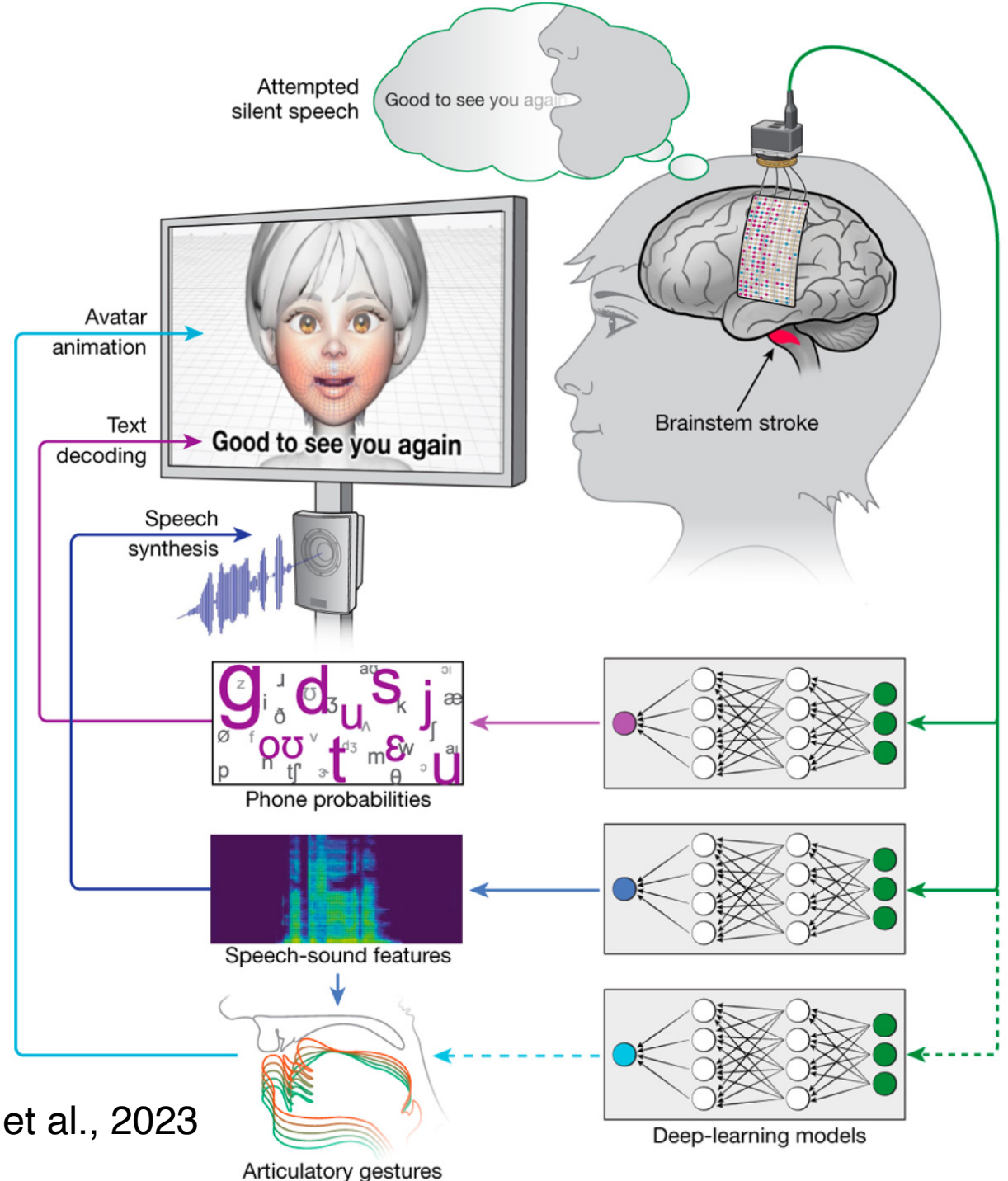
<https://doi.org/10.21203/rs.3.rs-3212709/v1>



Brain computer interfaces: for language

Multi-modal speech decoding.

- Neural activity was used to train a ANN to predict phone probabilities, speech-sound features
- and articulatory gestures.
- **A decoder was then constructed to produce text, generate audible speech and animate an avatar, respectively.**



Metzger et al., 2023

Brain computer interfaces: for visual scene estimation and reconstruction

Movie frame prediction from V1 of mice (using CEBRA + kNN decoder)

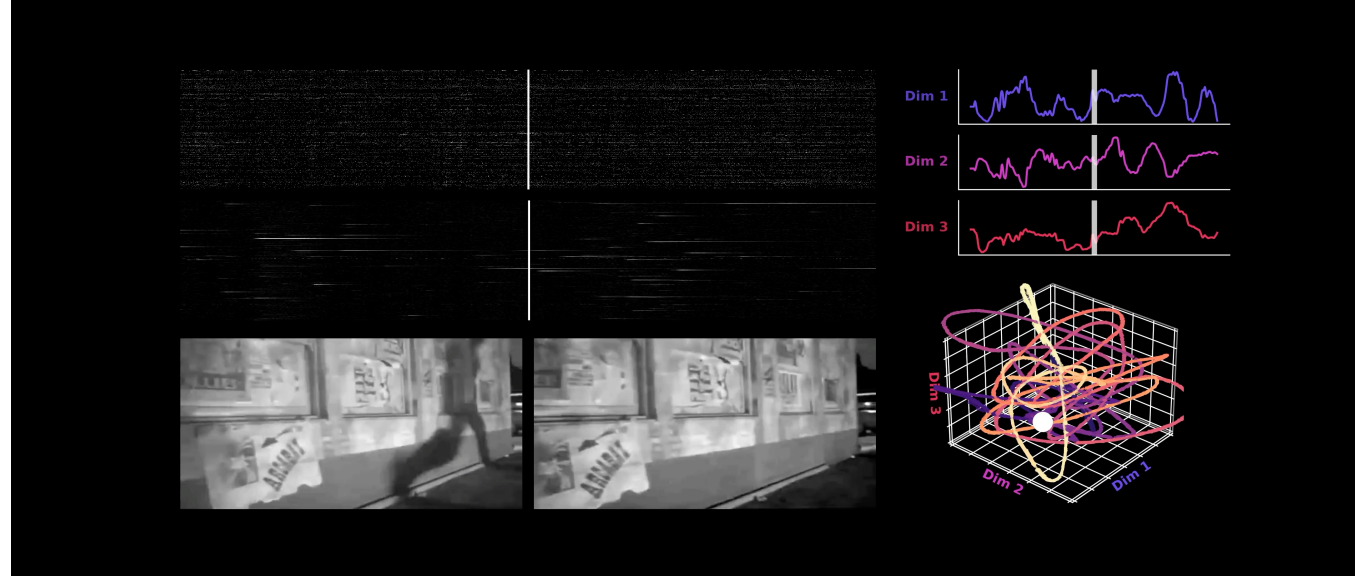


Illustration of decoded images from fMRI using Diffusion Models.

Ground truth (GT) vs. decoded images generated by Chen et al. from human fMRI with a diffusion model. Note that decoded images share similar color, shape and semantics.



What's next: BCIs also can be used to study neural dynamics of learning

BCIs not only have the potential to replace or augment motor function, **but also to be used as tools to study the direct and indirect neural circuits involved in learning as they adapt to new contingencies**

Garcia-Garcia et al. 2020

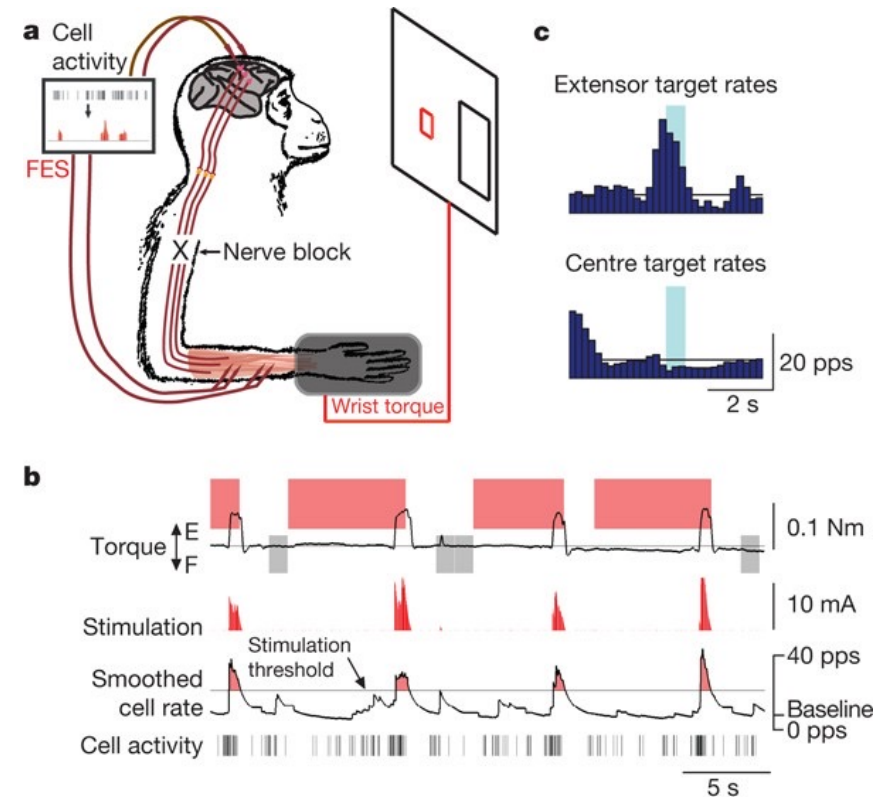
- Single neuron BCI!
- Monkeys demonstrated volitional control of the discharge rates of nearly all cells tested within the first 10-min practice session

Letter | Published: 01 December 2008

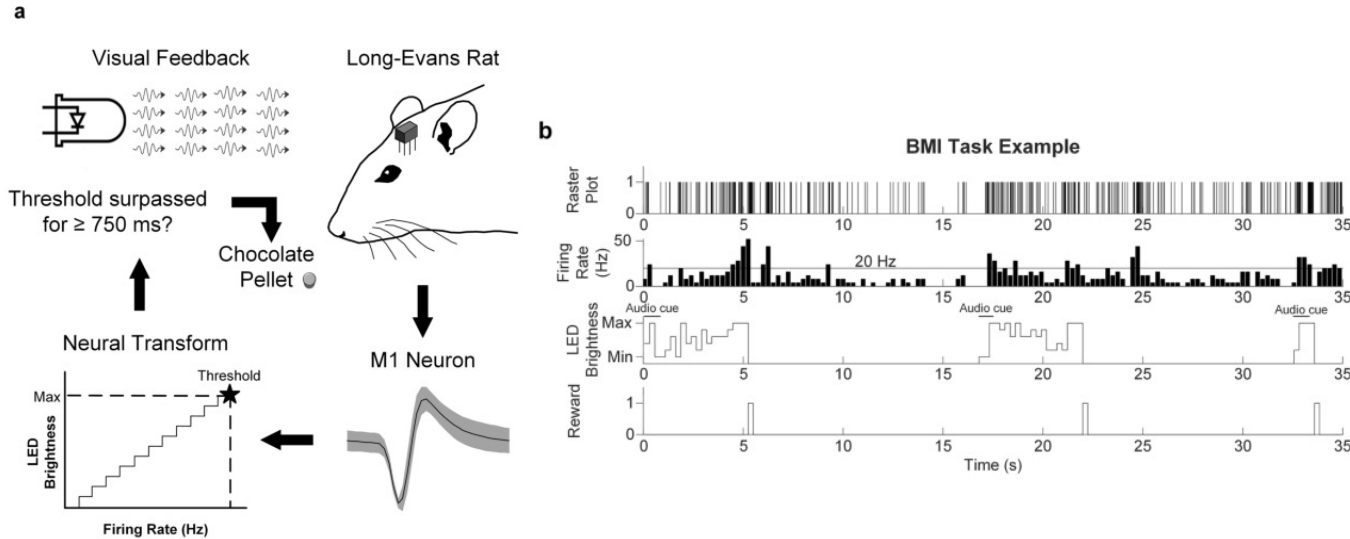
Direct control of paralysed muscles by cortical neurons

[Chet T. Moritz](#) , [Steve I. Perlmutter](#) & [Eberhard E. Fetz](#)

[Nature](#) 456, 639–642 (2008) | [Cite this article](#)



Brain computer interfaces: considering neuron subtypes



The authors investigated how neuron subtypes respond and adapt to a given BMI task.

- We conditioned single cortical neurons in a BMI task.
- Recorded neurons were classified into bursting and non-bursting subtypes based on their spike-train autocorrelation.
- Both neuron subtypes had similar improvement in performance and change in average firing rate.
- However, in bursting neurons, the activity leading up to a reward increased progressively throughout conditioning, while the response of non-bursting neurons did not change during conditioning.

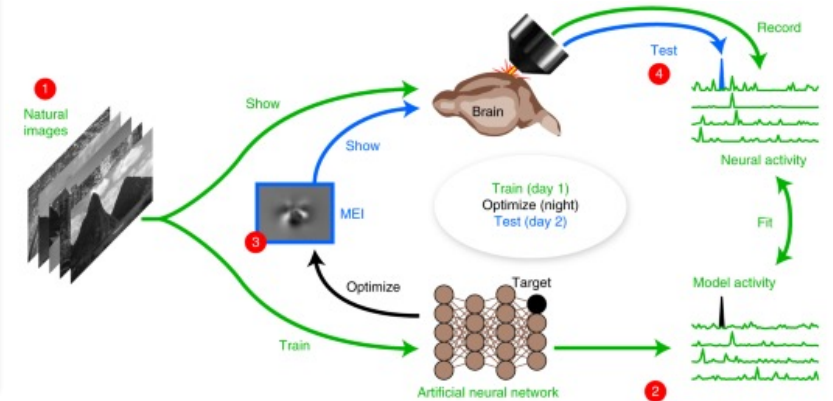
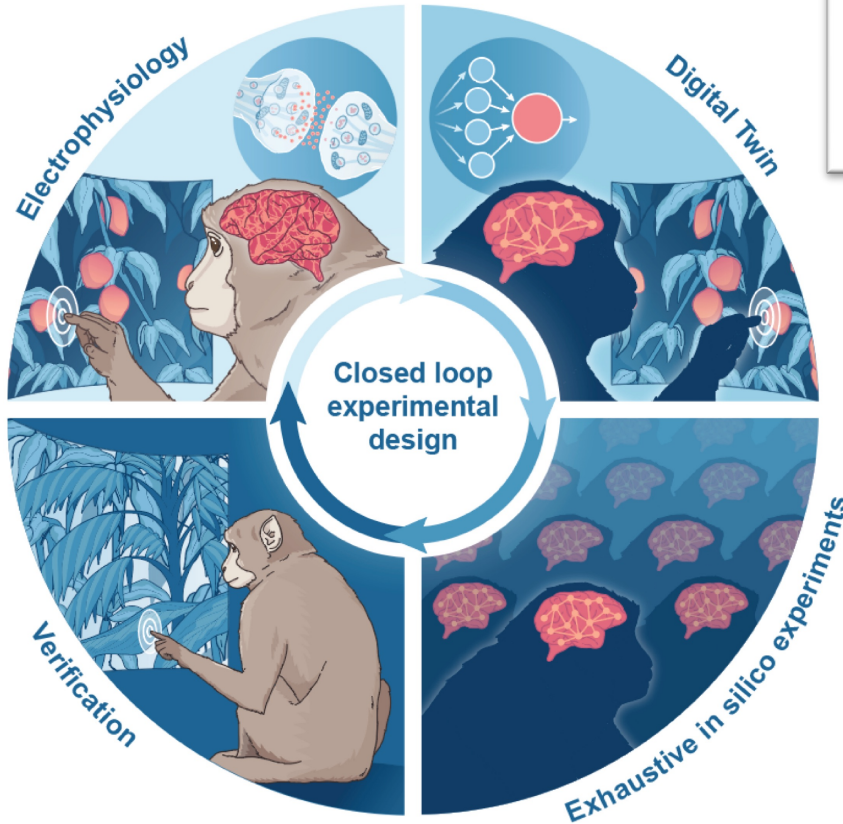
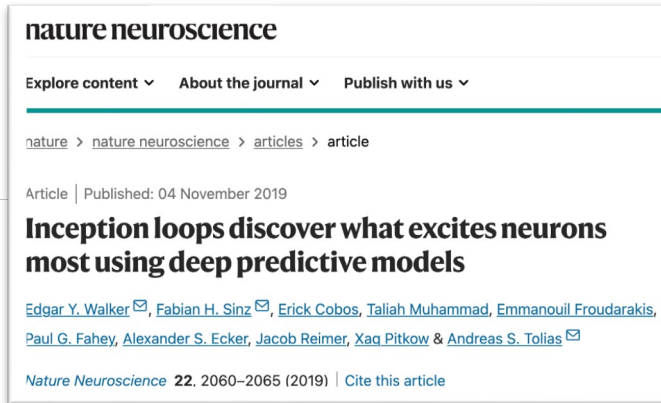
Article | [Open access](#) | Published: 17 November 2020

Operant conditioning of motor cortex neurons reveals neuron-subtype-specific responses in a brain-machine interface task

[Martha Gabriela Garcia-Garcia](#) , [Cesar Marquez-Chin](#) & [Milos R. Popovic](#)

[Scientific Reports](#) **10**, Article number: 19992 (2020) | [Cite this article](#)

Record neural data → build a model → deploy



- Inception Loops are a great example of neural decoding and close-loop stimulus design.
- In part, this is a type of “BCI”, as we are designing neural control of “optimal” external stimuli to drive activity.
- However, the goal of BCIs is slightly different: we want to “read & write” into the brain.

What's next: Reading & Writing into the brain

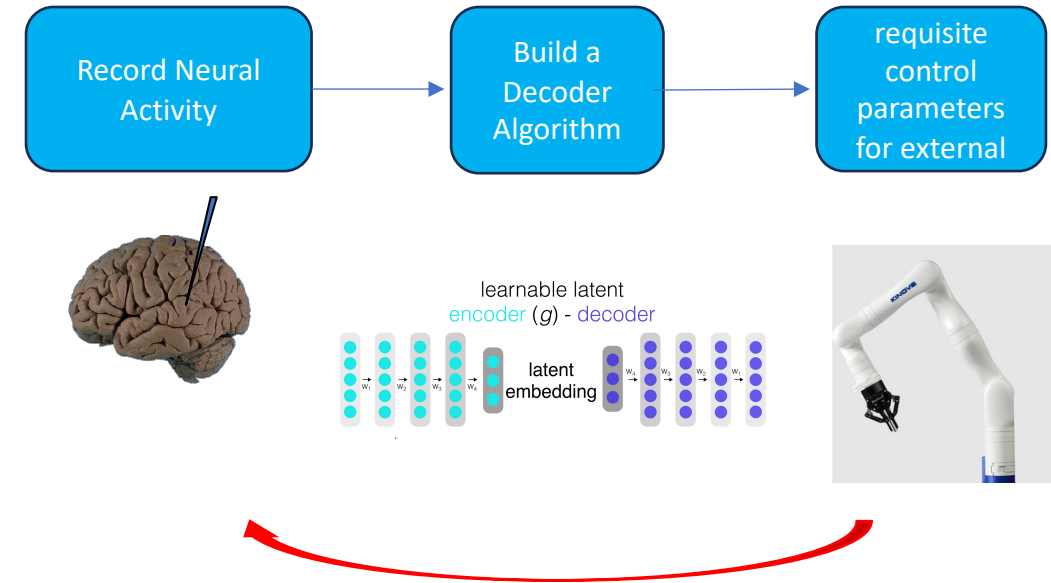
Current BCIs are mostly focused on allowing brain signals (neural activity) to be translated into controlling external devices. However, another critical goal is to give sensory feedback to the patient.

(imagine: if you are controlling a robotic arm, it would be very helpful to know when the arm touched the cup you want to pick up!)

This requires us then being to **DECODE** the signals to create the **MOVEMENTS** in the external device, and **GIVE SENSORY** feedback to the **BRAIN**.

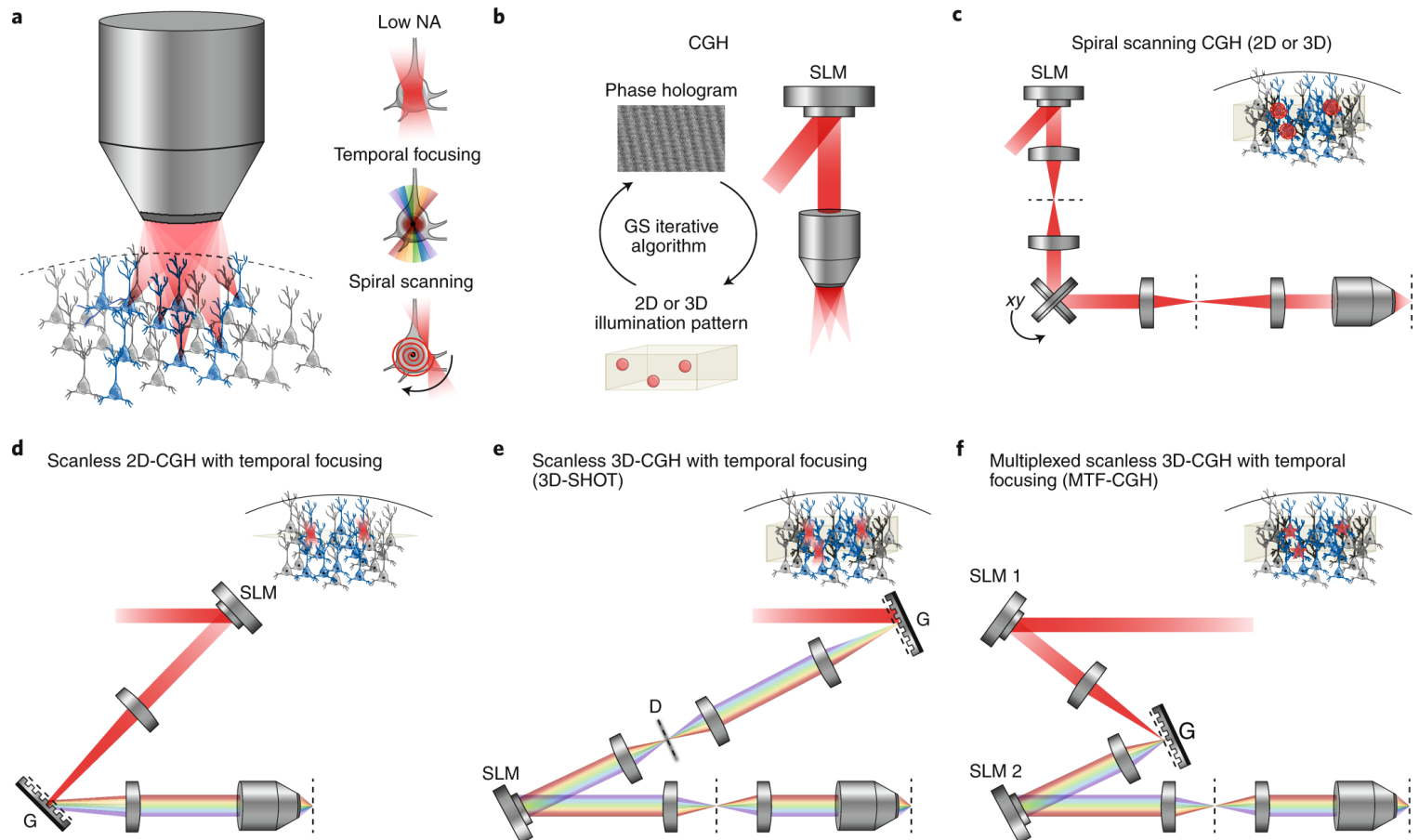
Do we know how to give the “right” patterns of activity to make it “feel, look, hear” real?

This is also where more scientific understanding is critical ...



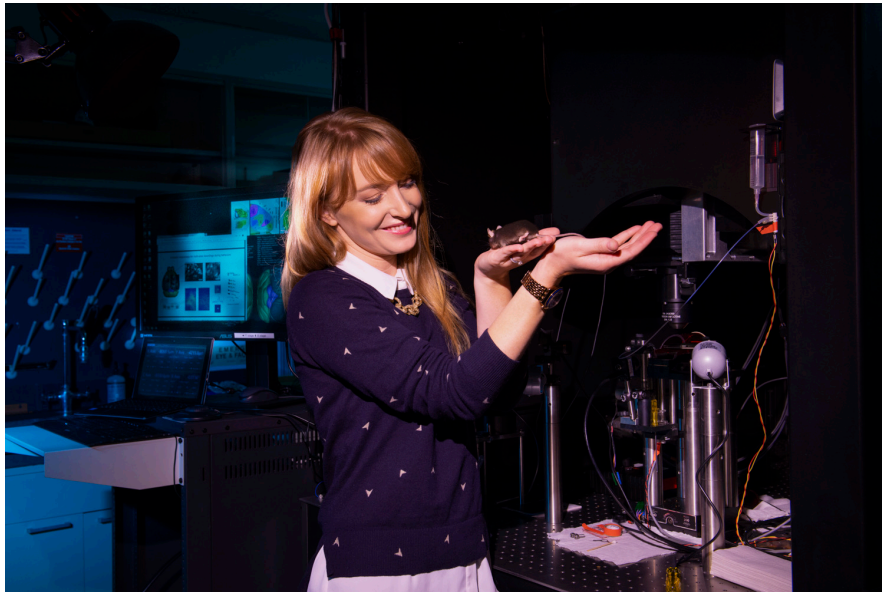
Reading & Writing into the brain: all optical studies

- **Calcium imaging + optogenetics** allows for “all optical” access the neural circuits.
- We can design closed-loop experiments to measure and perturb neural activity.
- We can design these such that we “closed-loop” record neural activity and have the animal use this activity to complete a task.

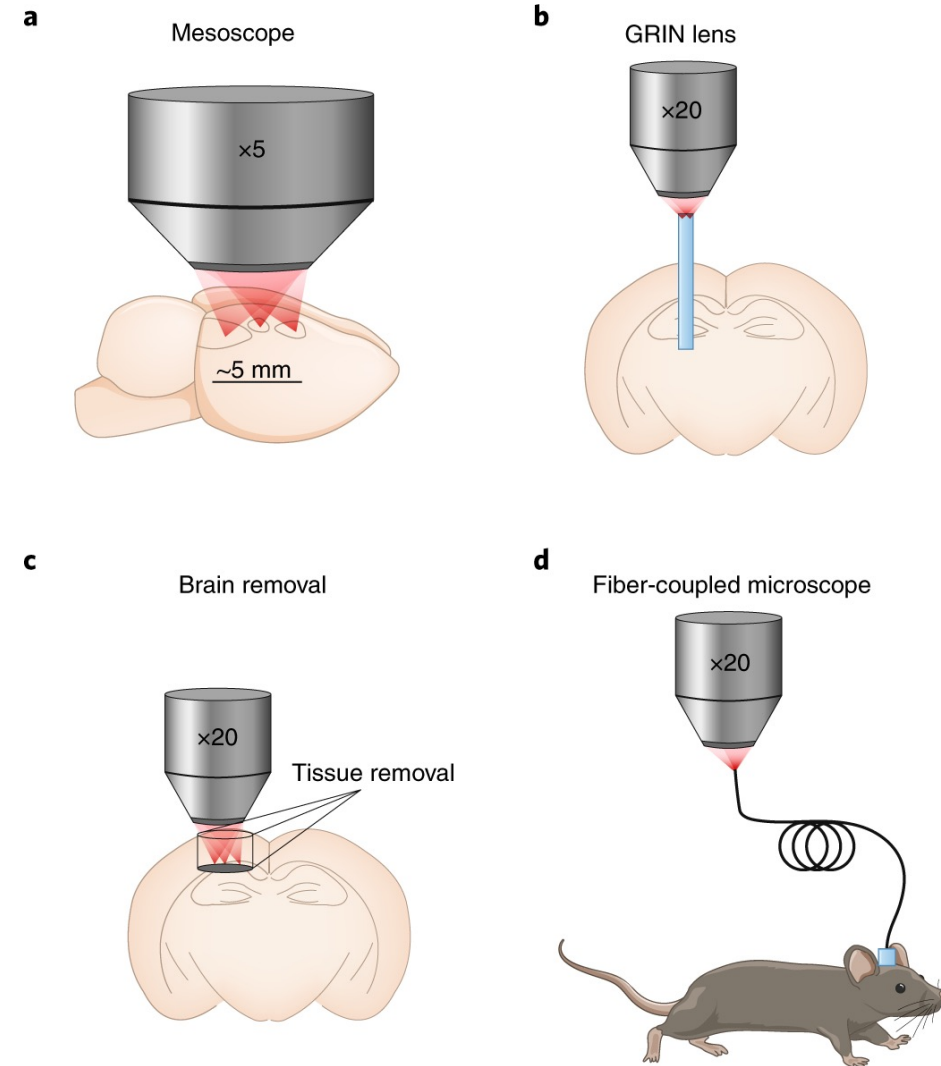


Reading & Writing into the brain: all optical studies

- Typical 2-photon imaging window is ~500 μm in diameter in the cortex (layers ~2-5 most typical).
- Now, new approaches are enabling recording in deeper areas (GRIN lens) and larger areas (mesoscope)

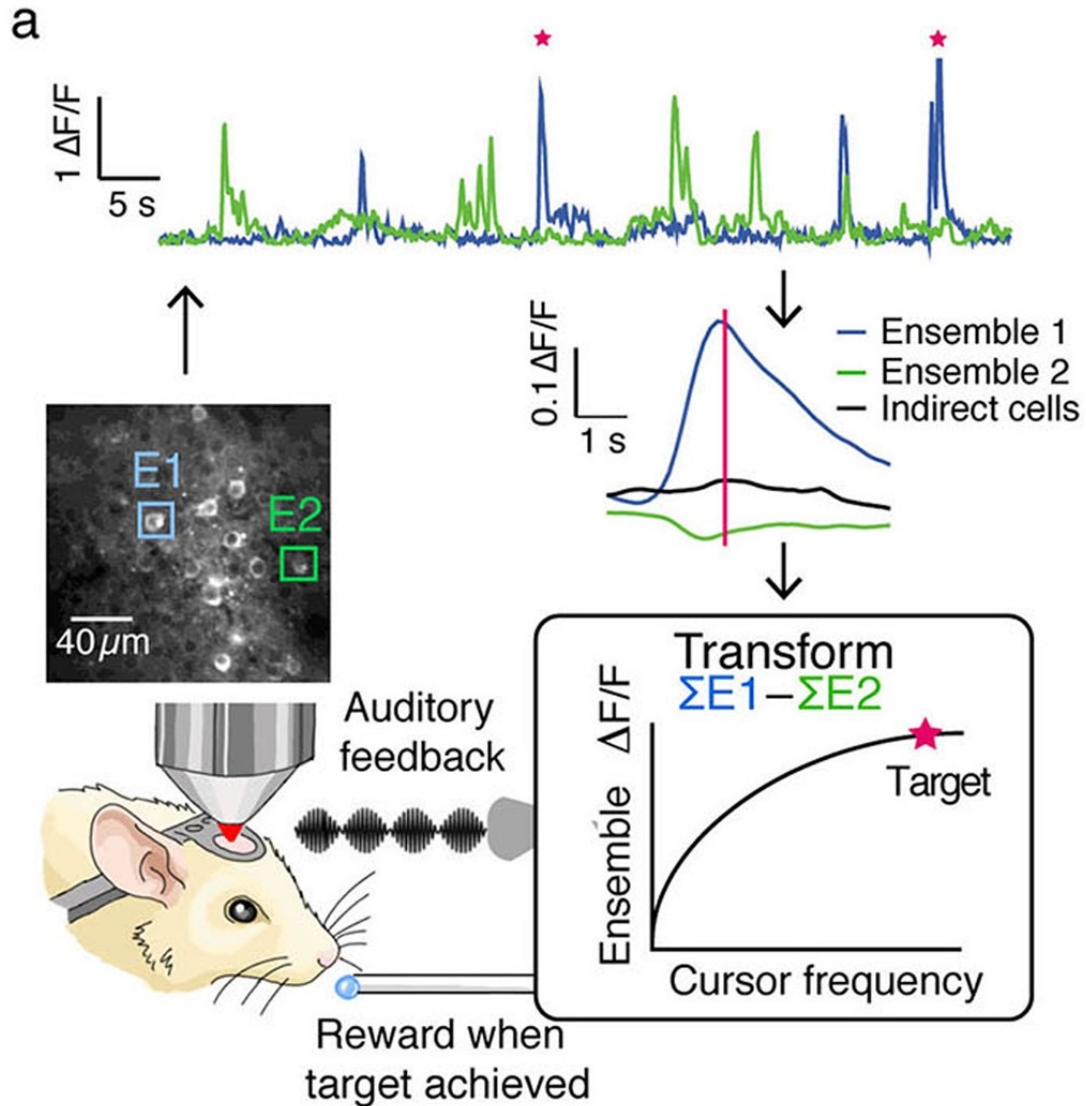


Our 2P mesoscope system at **EPFL**!
(not many in the world)



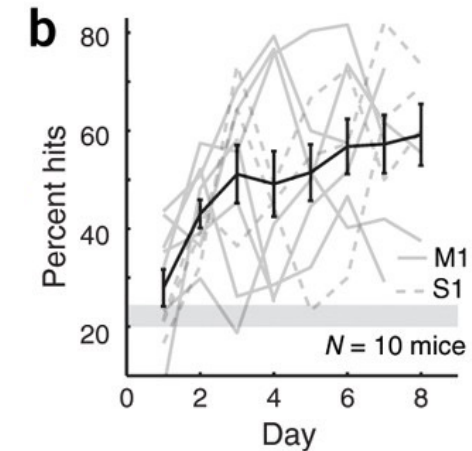
Adesnik, H., Abdeladim, L. Probing neural codes with two-photon holographic optogenetics. *Nat Neurosci* **24**, 1356–1366 (2021). <https://doi.org/10.1038/s41593-021-00902-9>

Brain computer interfaces: all optical studies

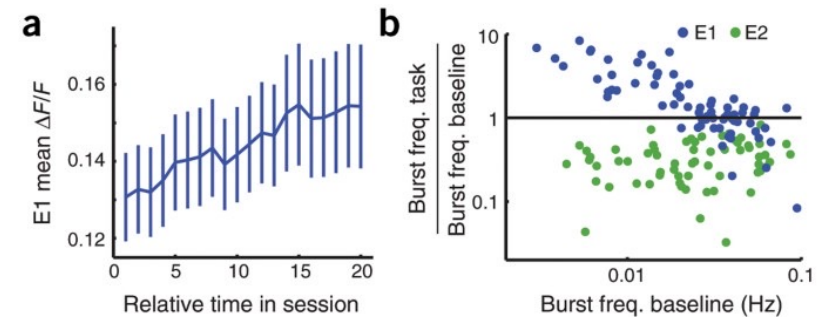


BMI tasks provide a powerful approach for studying sensorimotor learning, as they enable arbitrary mapping between neuronal activity, behavioral output and reward

Mice learn to intentionally modulate calcium dynamics.



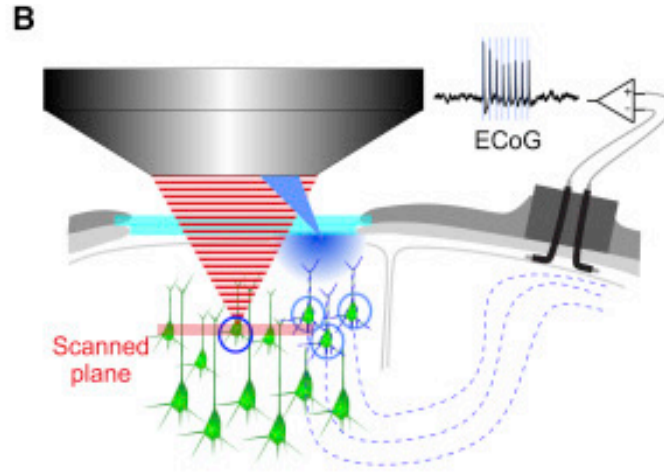
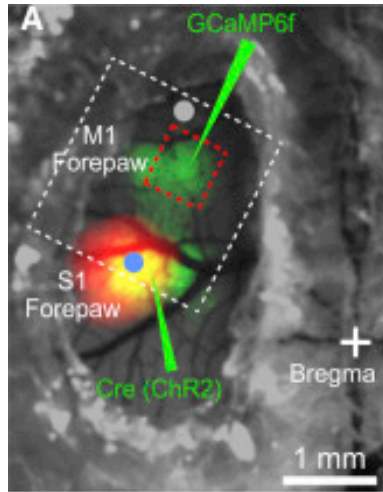
Local network reorganization accompanies learning



Brain computer interfaces: all optical studies

Can mice learn to use a fabricated feedback channel to control single-neuron activity?

How do responses of the conditioned and neighboring cortical neurons change with learning?

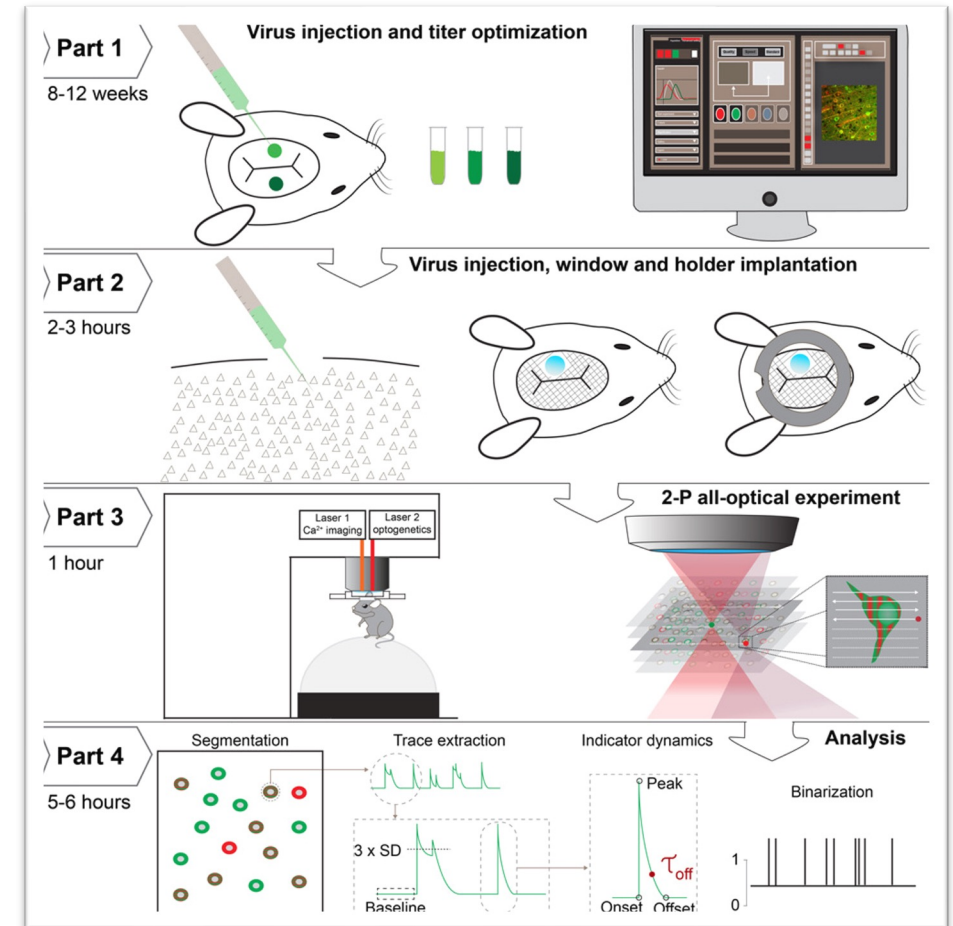


- They used 2P imaging, optogenetics, ECoG, real-time ROI selection, and a decoding algorithm*
- (*technical tour de force!*)

*general linear model with the threshold crossing rate (TCR) z score as the dependent and the z scores of the event rate (ER) and event amplitude (EA) as the independent variables was fit to each of the sessions as follows:

$$TCR = b_{Er}ER + b_{Ea}EA$$

Reminder: 2P imaging

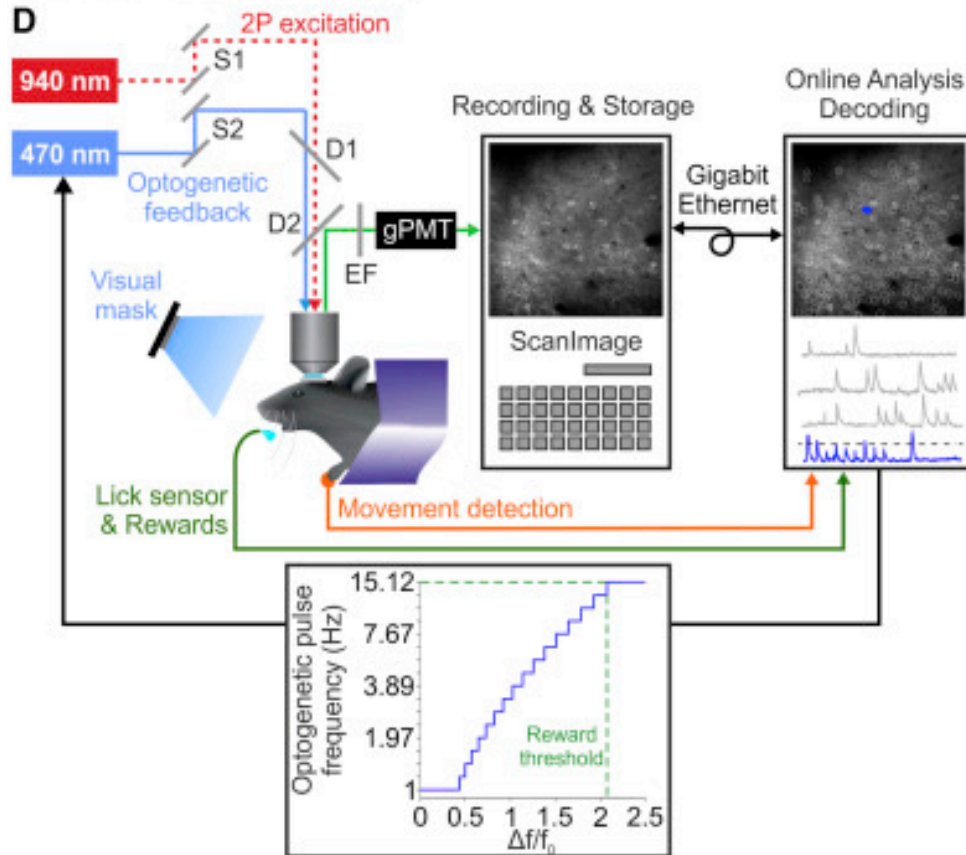


<https://www.sciencedirect.com/science/article/pii/S2666166721007164>

Brain computer interfaces: all optical studies

Can mice learn to use a fabricated feedback channel to control single-neuron activity?

How do responses of the conditioned and neighboring cortical neurons change with learning?



Schematic of the experimental setup showing the head-fixed mouse under the two-photon microscope.

The two laser beams for imaging and stimulation were controlled independently, but focused through the same microscope objective.

To protect the sensitive photomultiplier tubes of the imaging system during the optogenetic blue light pulses, we combined optical filters and gated detection electronics.

Two-photon images were acquired on a PC running Scanimage and streamed to a control PC that extracted in real time:

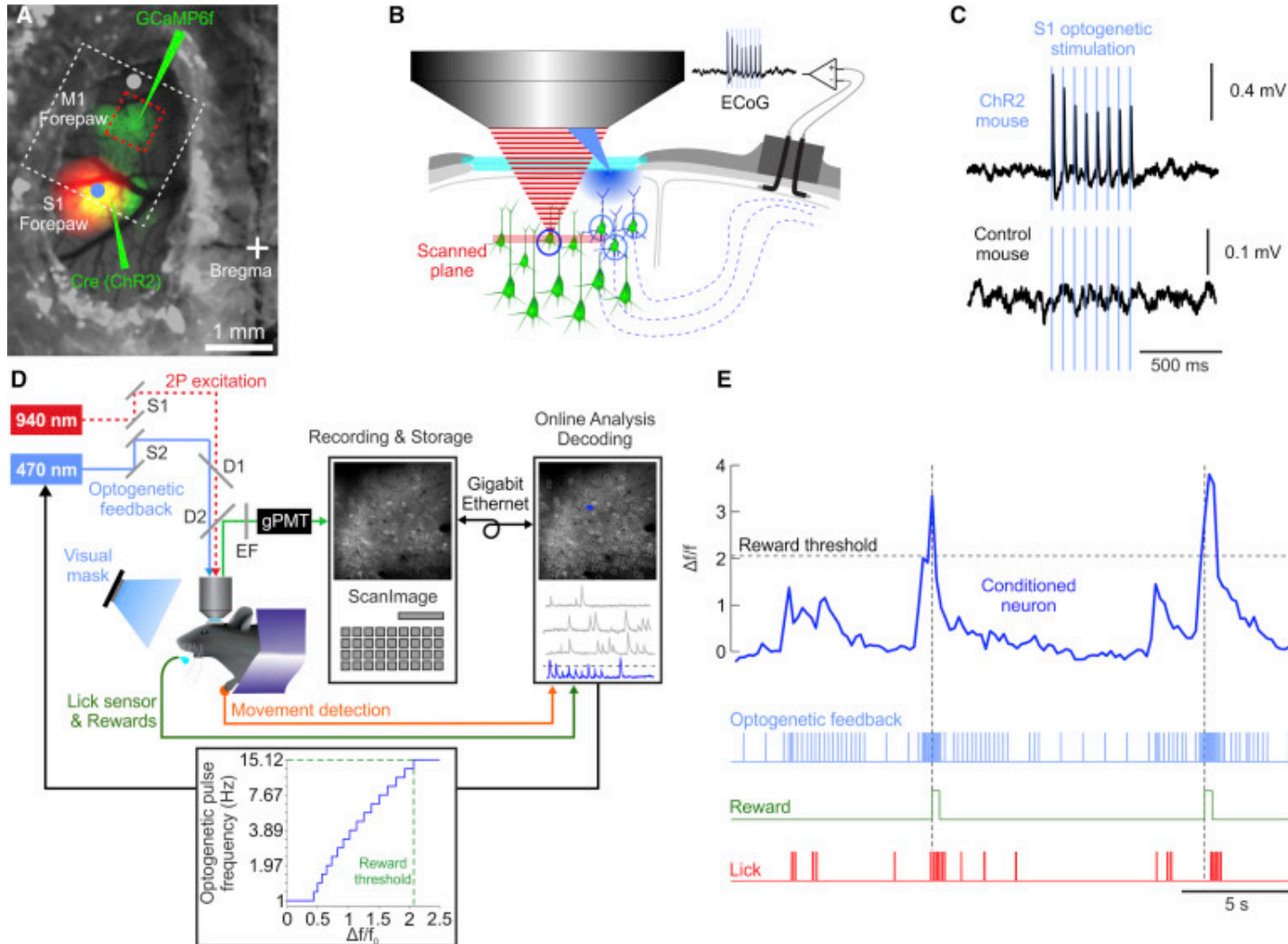
- the Ca-dependent neural activity
- generated the feedback signal (inset: transfer function between neural activity and feedback pulse rate)
- acquired ECoG, lick, and forepaw movement sensor data
- controlled the blue light mask

S1, S2, scanning mirrors; D1, D2, dichroic mirrors; EF, emission filter; gPMT, gated photomultiplier tube. A conditioned neuron (CN) (blue) was chosen among hundreds of simultaneously imaged neurons (gray).

Brain computer interfaces: all optical studies

Can mice learn to use a fabricated feedback channel to control single-neuron activity?

How do responses of the conditioned and neighboring cortical neurons change with learning?



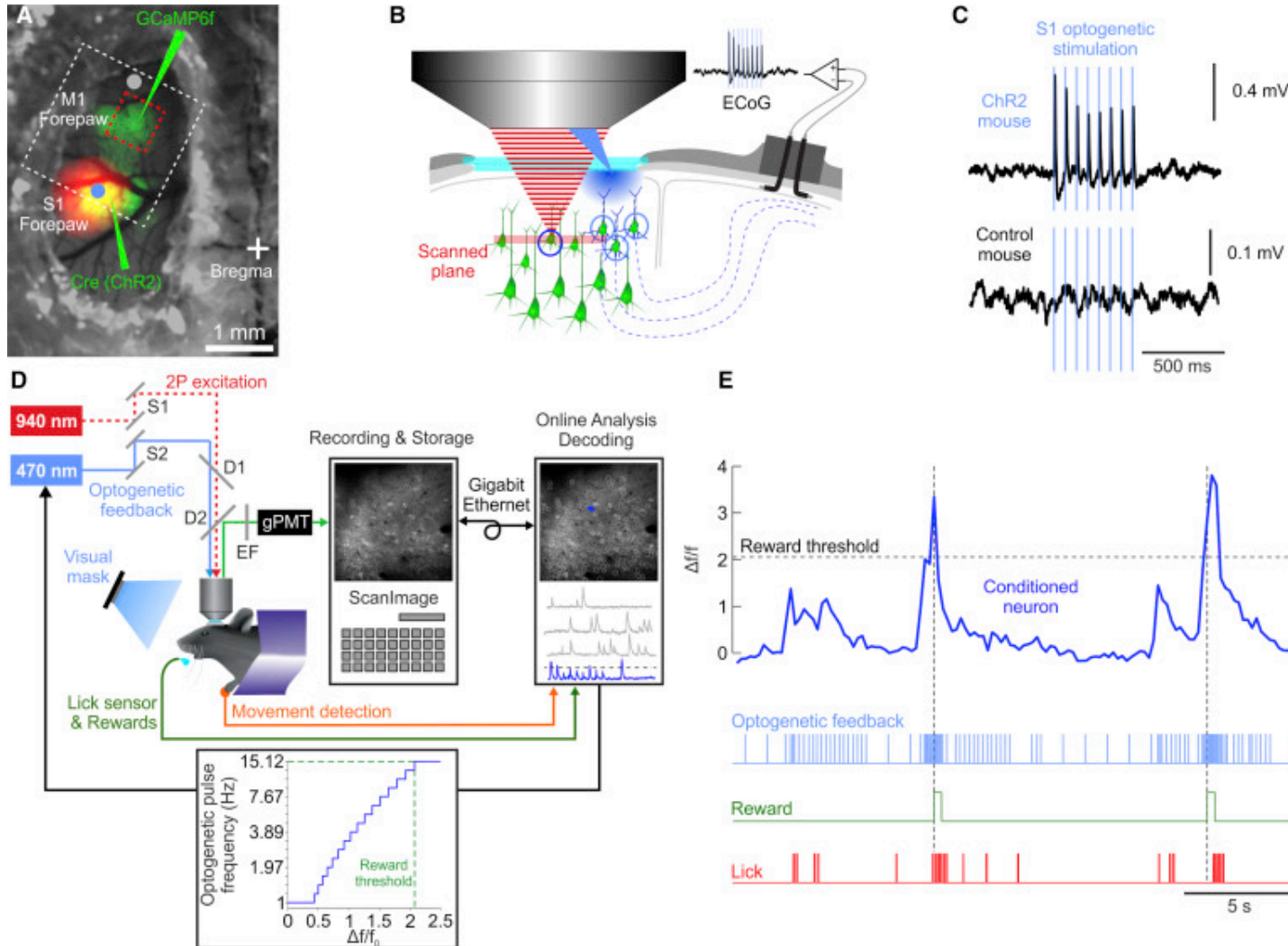
- All-optical brain-machine-brain interface for neuroprosthetic control
- Mice rapidly learn to activate single neurons under optogenetically evoked feedback
- Population imaging reveals that learning is restricted to the conditioned neuron

Note, no visual feedback was provided, only stimulation into S1!

Brain computer interfaces: all optical studies

Can mice learn to use a fabricated feedback channel to control single-neuron activity?

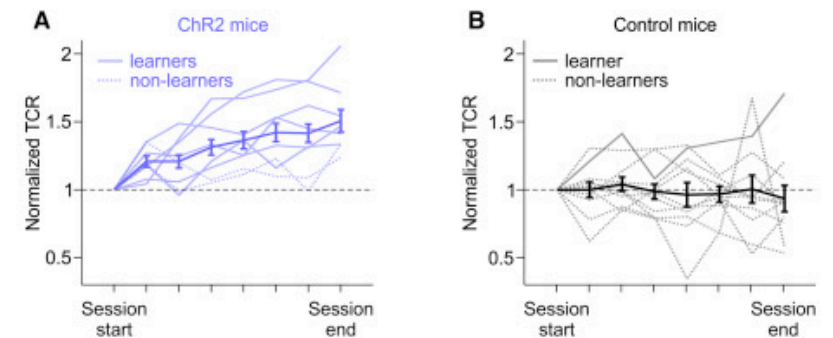
How do responses of the conditioned and neighboring cortical neurons change with learning?



- All-optical brain-machine-brain interface for neuroprosthetic control
- Mice rapidly learn to activate single neurons under optogenetically evoked feedback
- Population imaging reveals that learning is restricted to the conditioned neuron

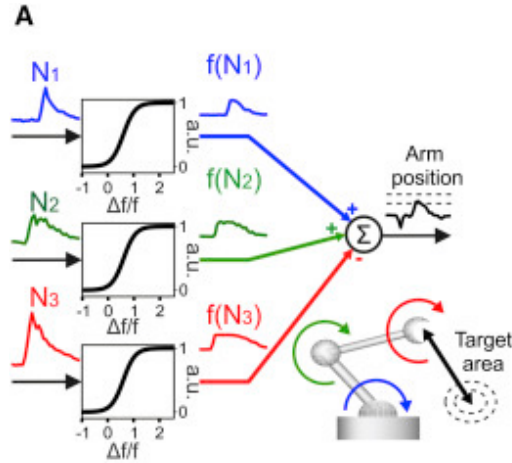
Note, no visual feedback was provided, only stimulation into S1:

Importance of Artificial Sensory Feedback for Execution and Learning

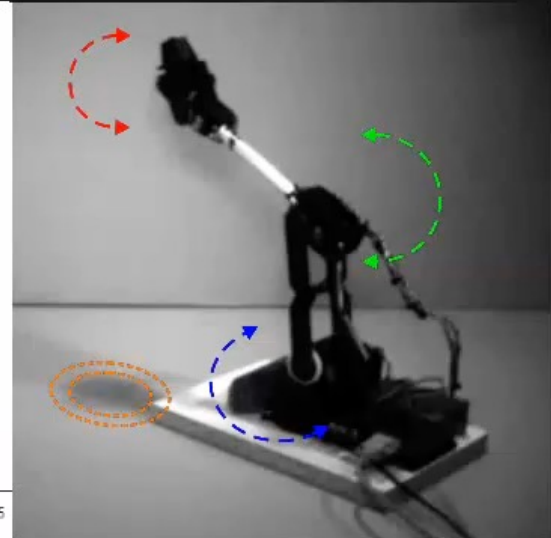
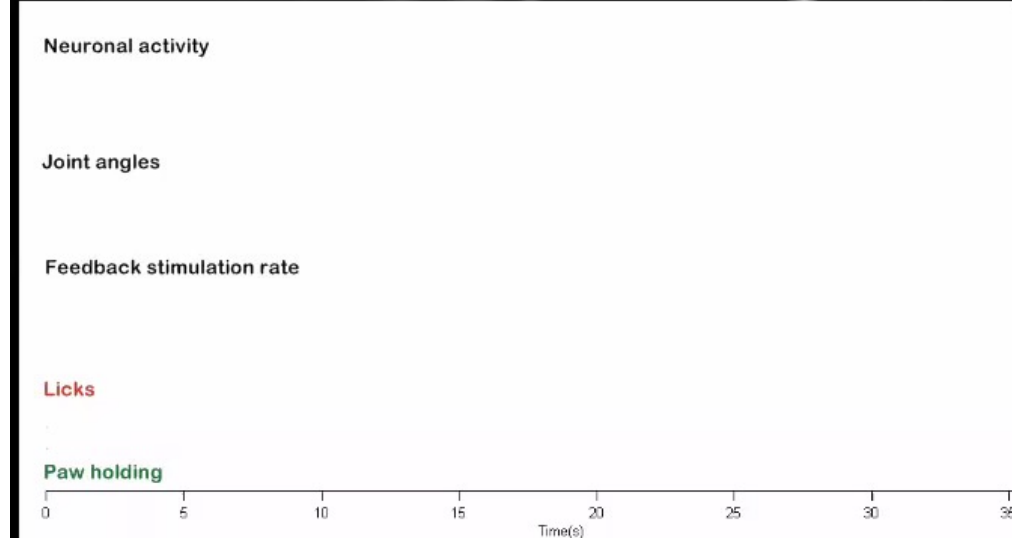
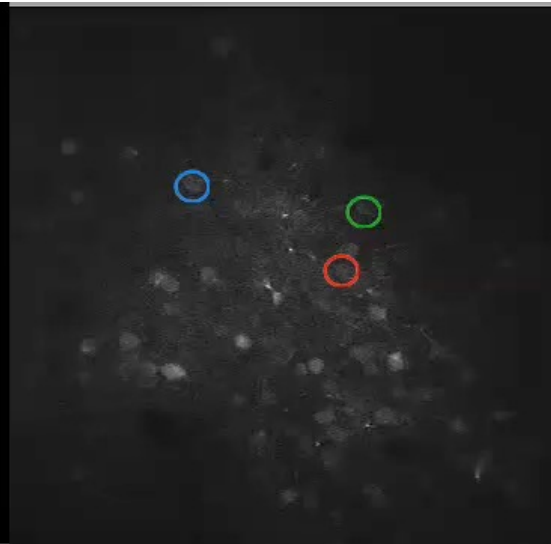
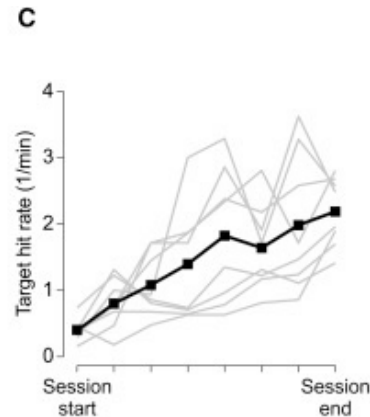
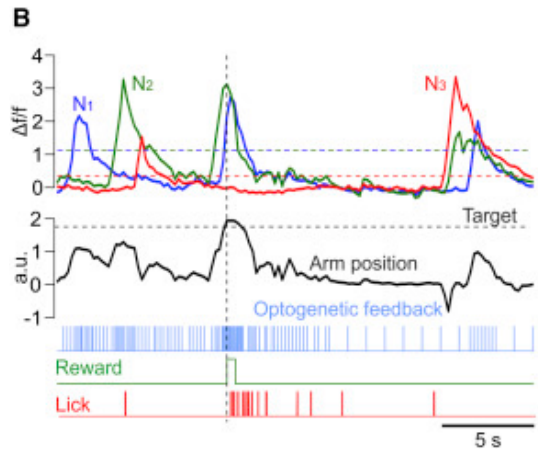


Brain computer interfaces: all optical studies

Multi-neuron Conditioning



The $\Delta f/f_0$ activities N_1 , N_2 , and N_3 of three CNs were first transformed by a logistic function into $f(N_1)$, $f(N_2)$, and $f(N_3)$, and the ensemble activity computed by adding $f(N_1)$ and $f(N_2)$ and subtracting $f(N_3)$. Each activity transform dictated the angular position of a robotic actuator and the ensemble activity was proportional to the effector distance to target.



Designing experiments to test neural coding

a, Probing the role of neural synchrony.

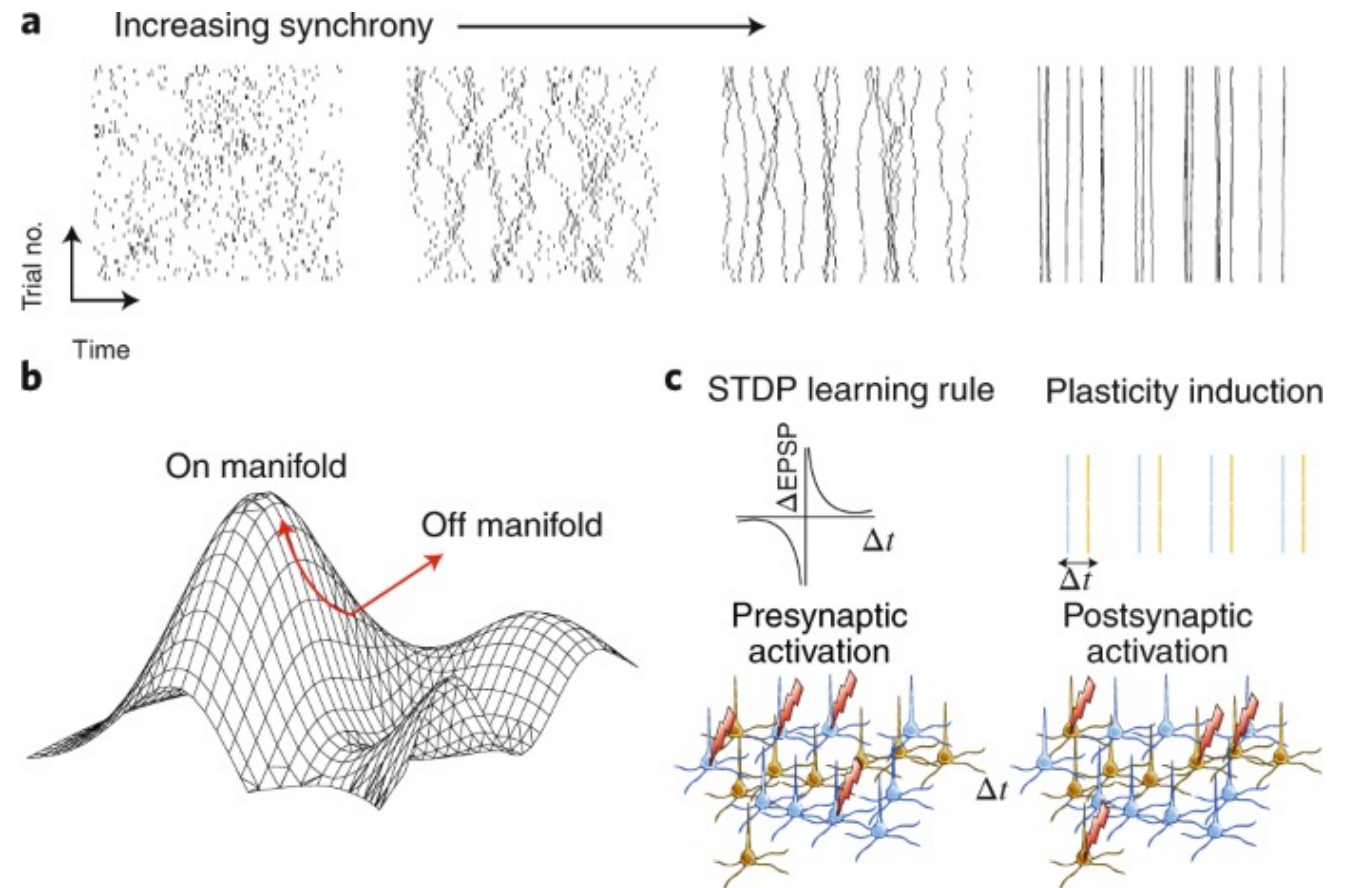
b, Schematic of two types of multiphoton holographic perturbations, one that obeys the intrinsic low-dimensional architecture of the neural activity patterns a group of neurons exhibits ('on manifold') and one that does not ('off manifold').

c, Schematic of using multiphoton holographic optogenetics to artificially induce Hebbian spike timing-dependent plasticity between two artificially chosen neural ensembles (indicated by the two colors).

Top left, schematic of the conventional STDP learning rule, plotting change in synaptic strength (ΔEPSP) versus the time delay between pre- and postsynaptic activation (t).

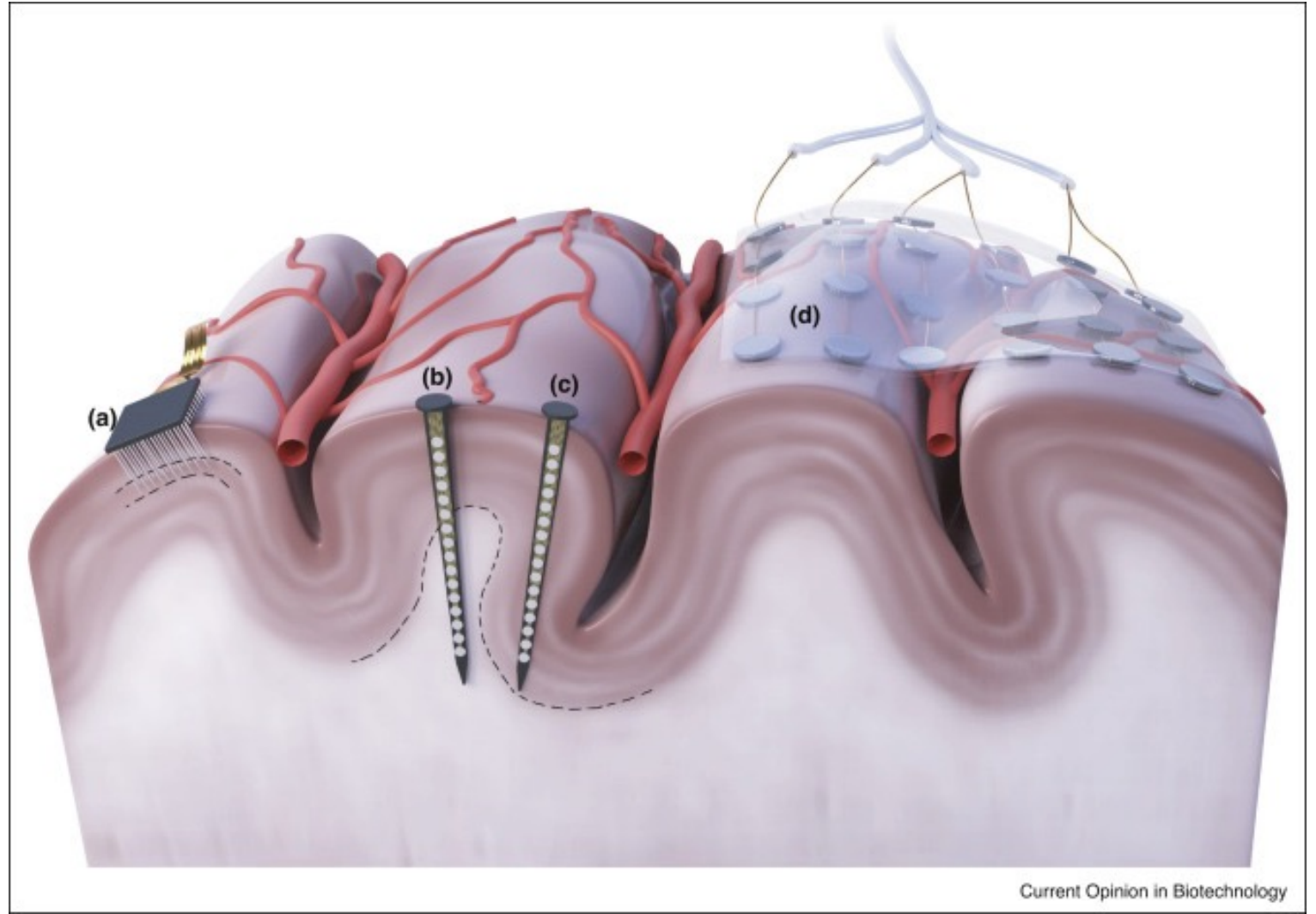
Top right, holographic stimulus pattern to induce STDP.

Bottom, schematic of fast interleaved photostimulation of two ensembles to drive STDP between them.



Brain computer interfaces: electrical implants

BMIs in humans and non-human primates are generally all electrical systems.

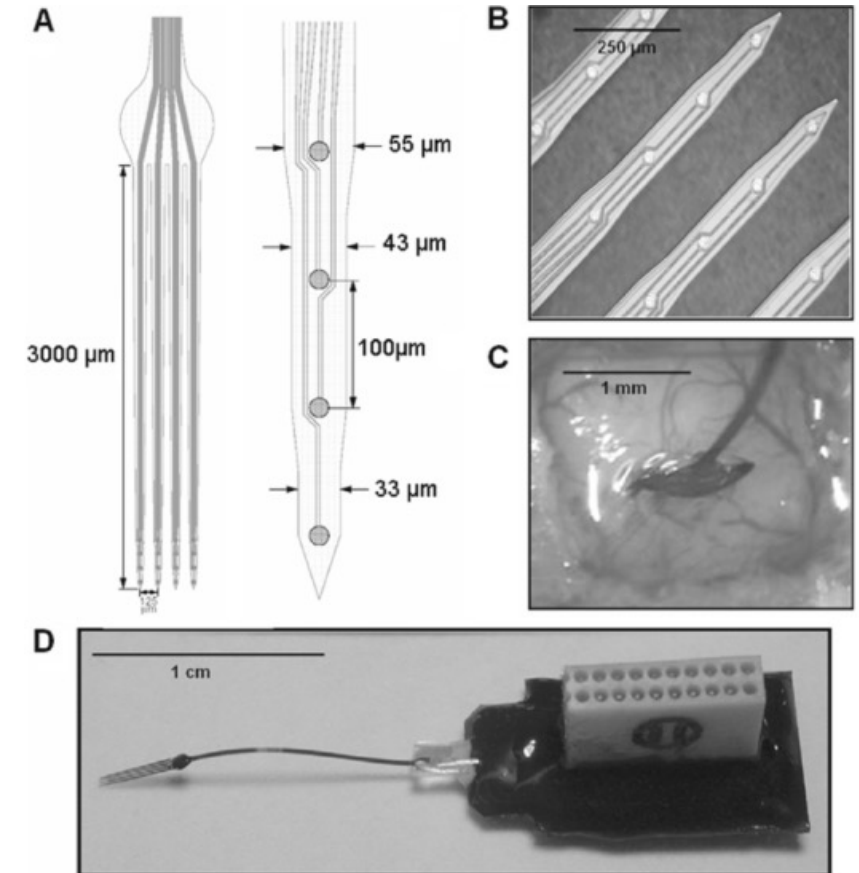
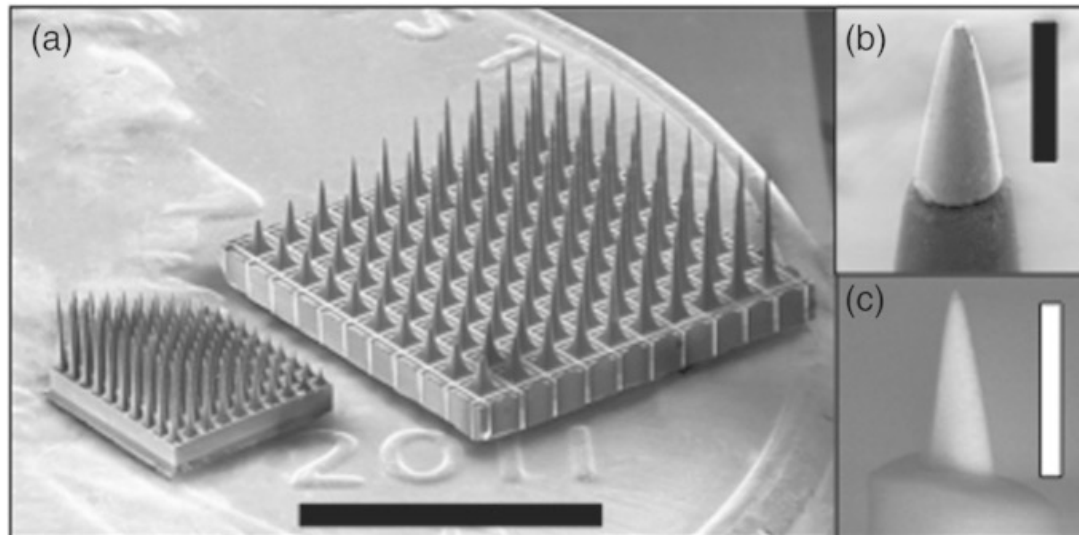


Rapeaux & Constandinou, 2021
Current opinion Biotech.

Implanted microelectrode arrays

Implanted microelectrode arrays are a common tool in systems neuroscience.

The size and number of recording sites vary with experimental goals and model organism.



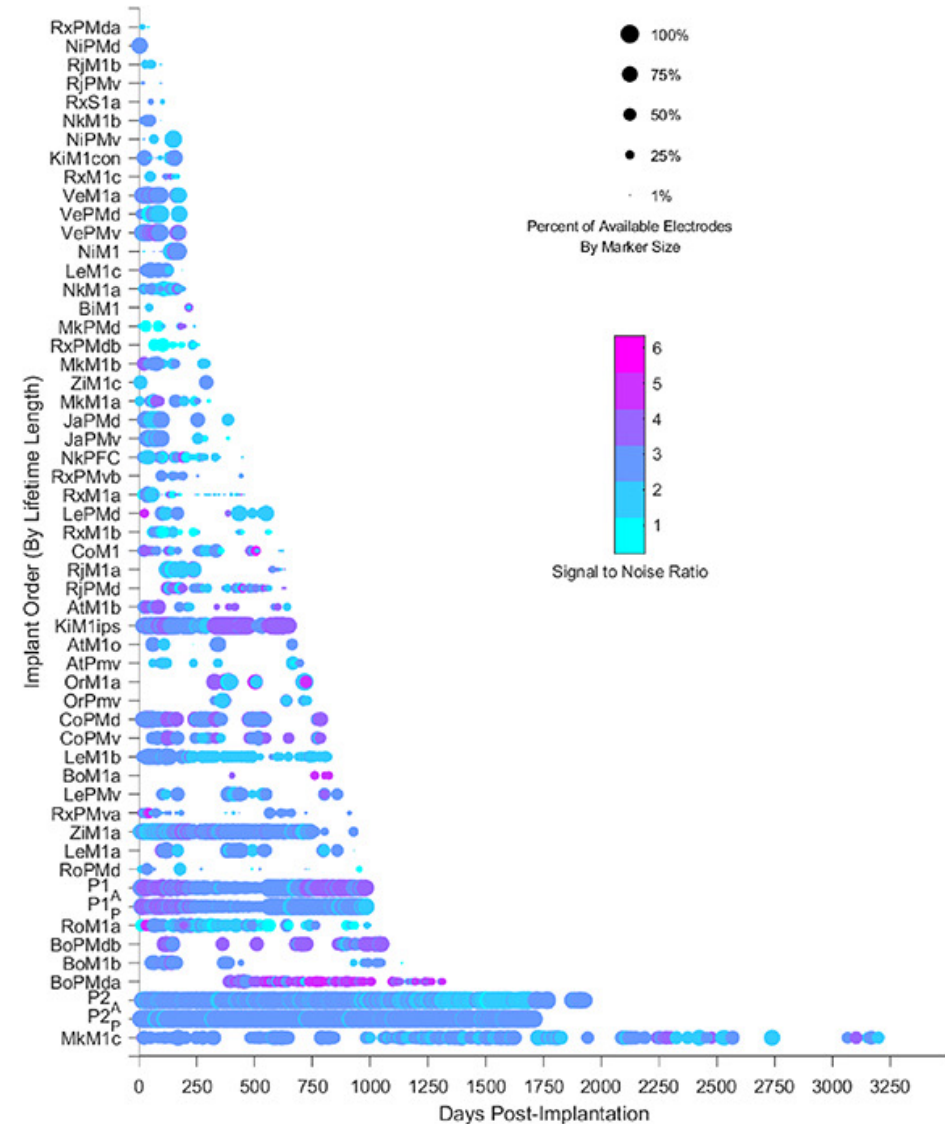
Vetter et al.,
2004
*Transactions on
Biomed. Eng.*

Wark et al., 2013 *J. Neural Eng.*

Microelectrode arrays

Implanted arrays can provide stable recordings for months or years.

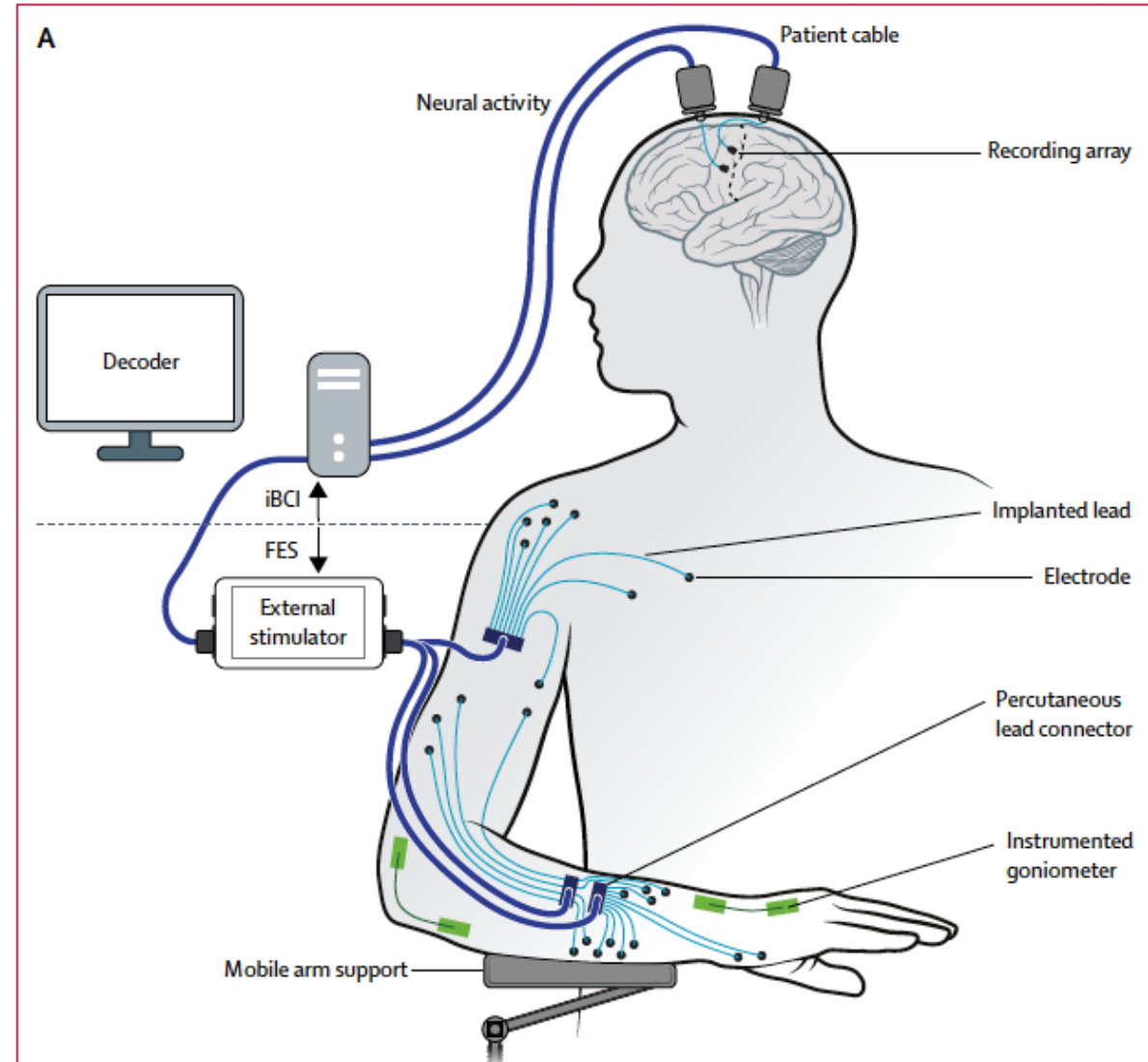
In primate research, research subjects often live with cortical implants for years, and perform multiple tasks with researchers.



Sponheim et al., 2021 *J. Neural Eng.*

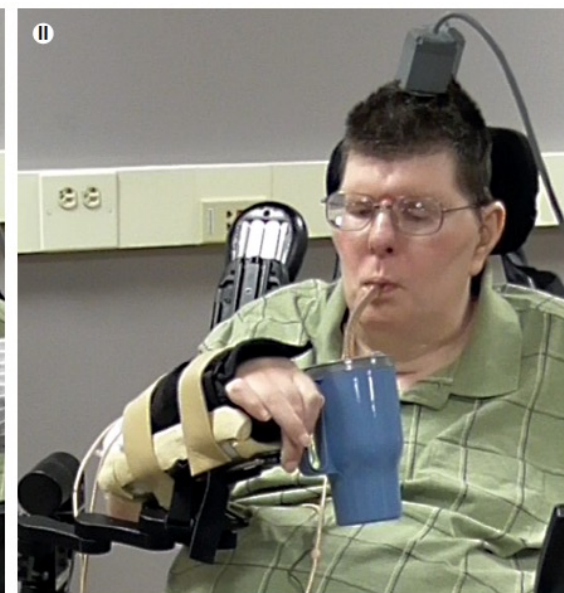
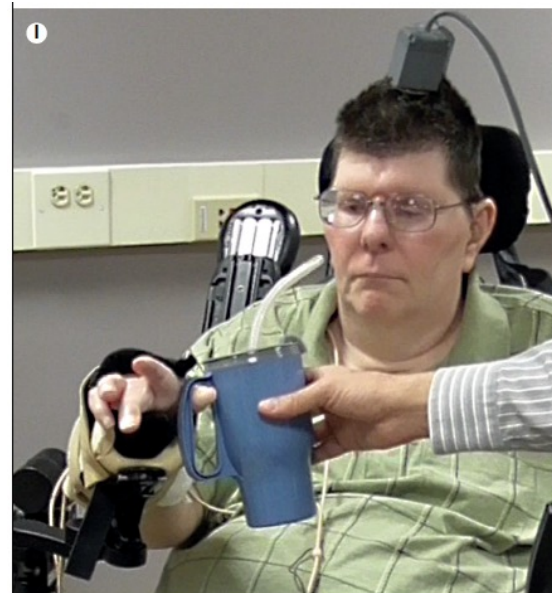
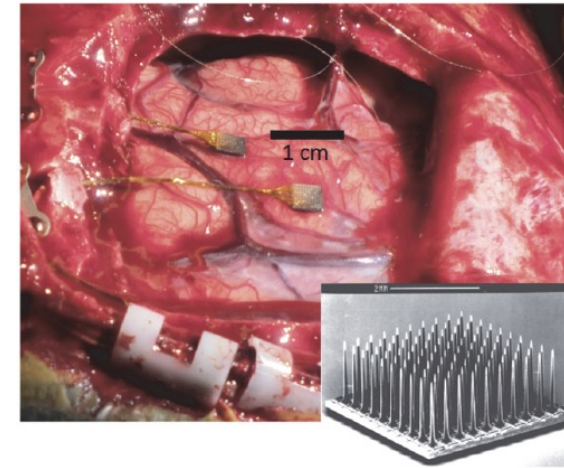
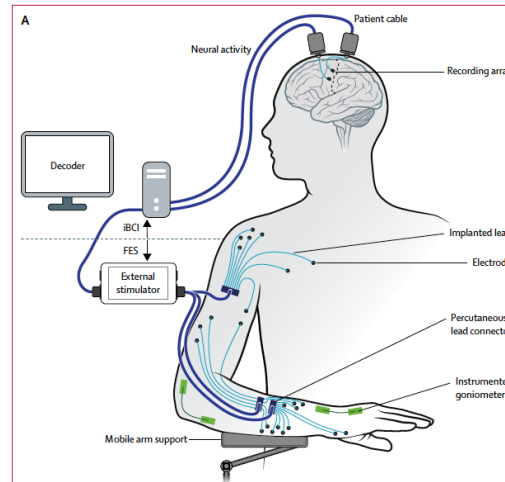
Brain computer interfaces: electrical implants in humans

- Implanted arrays have also been tested in clinical settings.
 - They have been used to help quadriplegics regain some lost function.



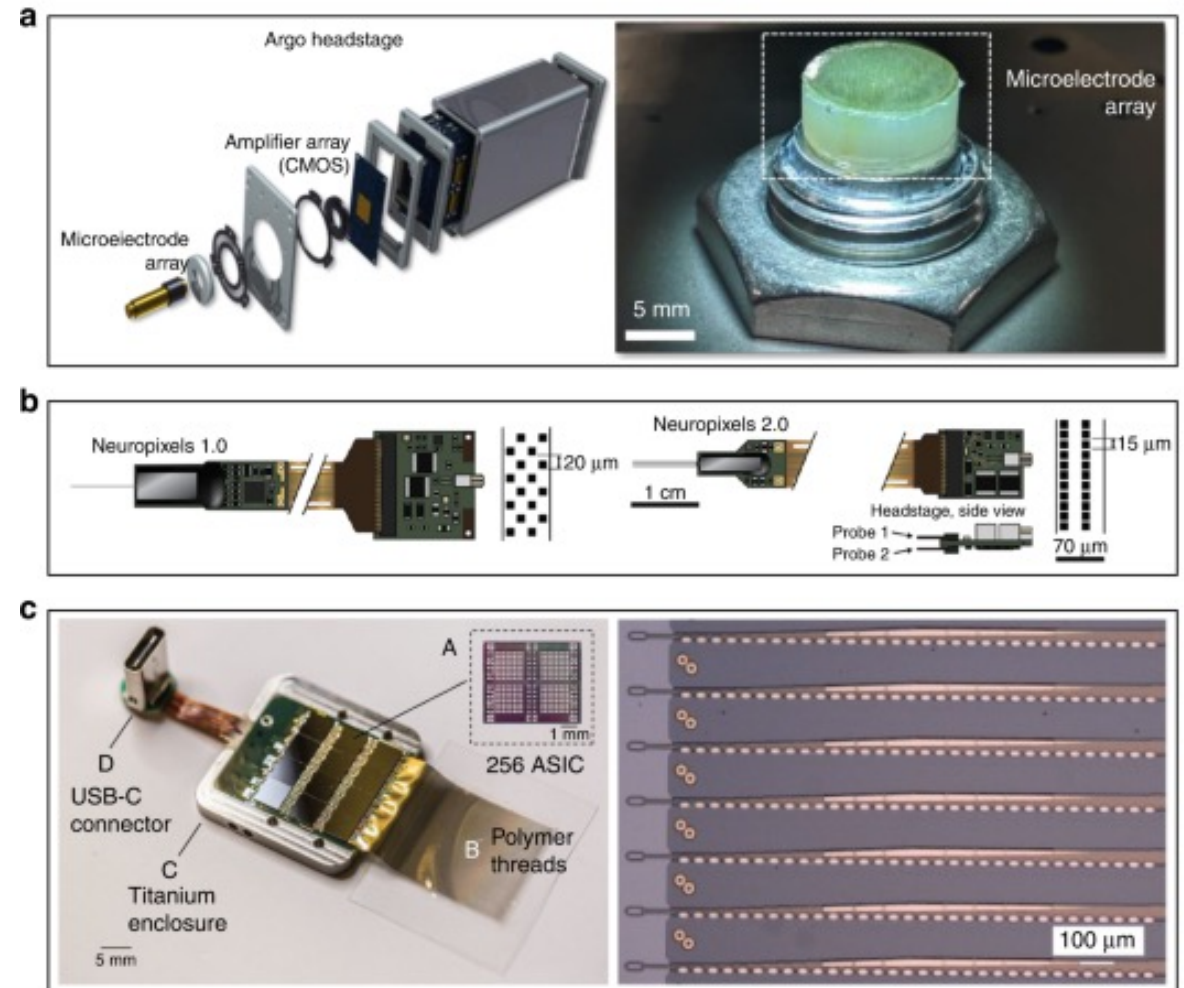
Brain computer interfaces: electrical implants in humans

- Implanted arrays have also been tested in clinical settings.
 - They have been used to help quadriplegics regain some lost function.
- However, the use of cortical BMIs is almost entirely experimental



Next gen microelectrode arrays

- Modern microelectrodes tend to have increased throughput.
 - 1000-6000 contacts
- There is also a focus on more biocompatible materials.
- The electrode arrays used in research are several years ahead of their clinical counterparts due to safety testing requirements.



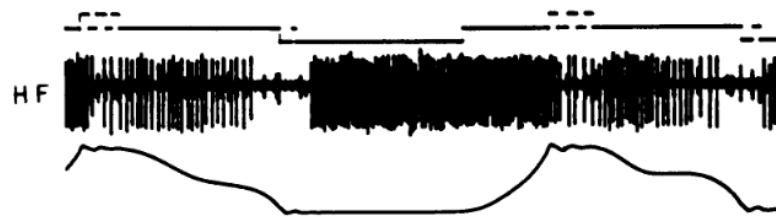
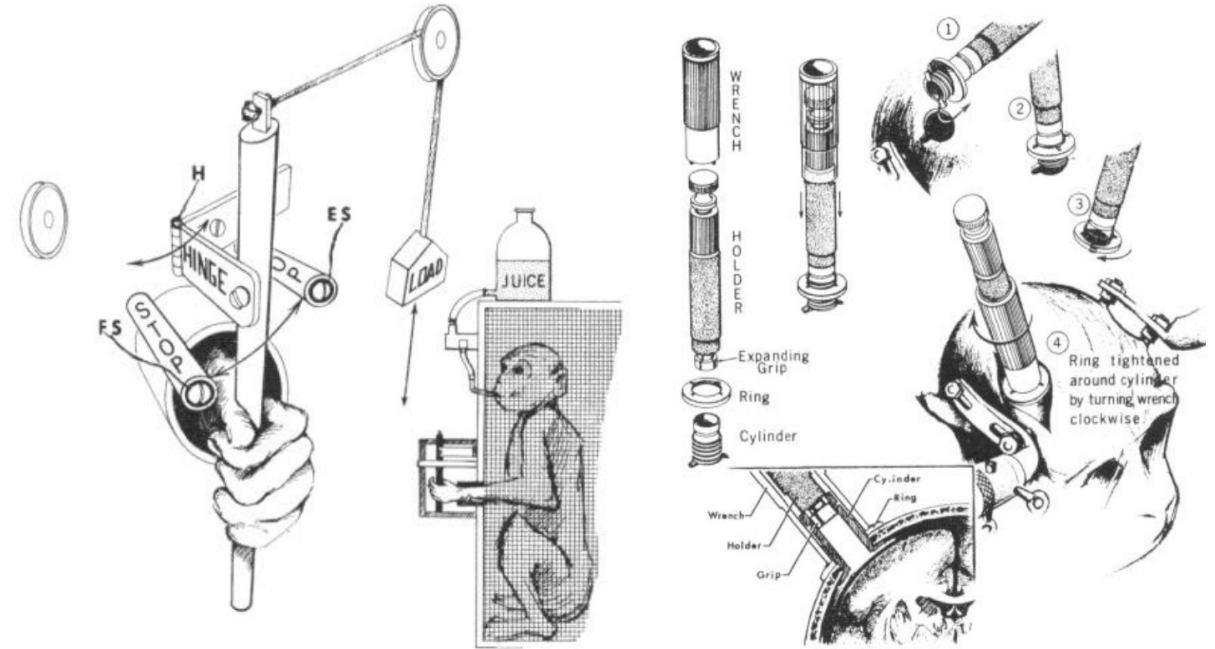
Wang et al., 2023 *Microsystems & NeuralEng.*

So we can record from motor cortex with BMIs, what have researchers found?

Movement research and primates

In vivo recordings of cortex in primates have been used to study movement for over 50 years.

Even the earliest studies demonstrated motor cortex encodes key features of movements.



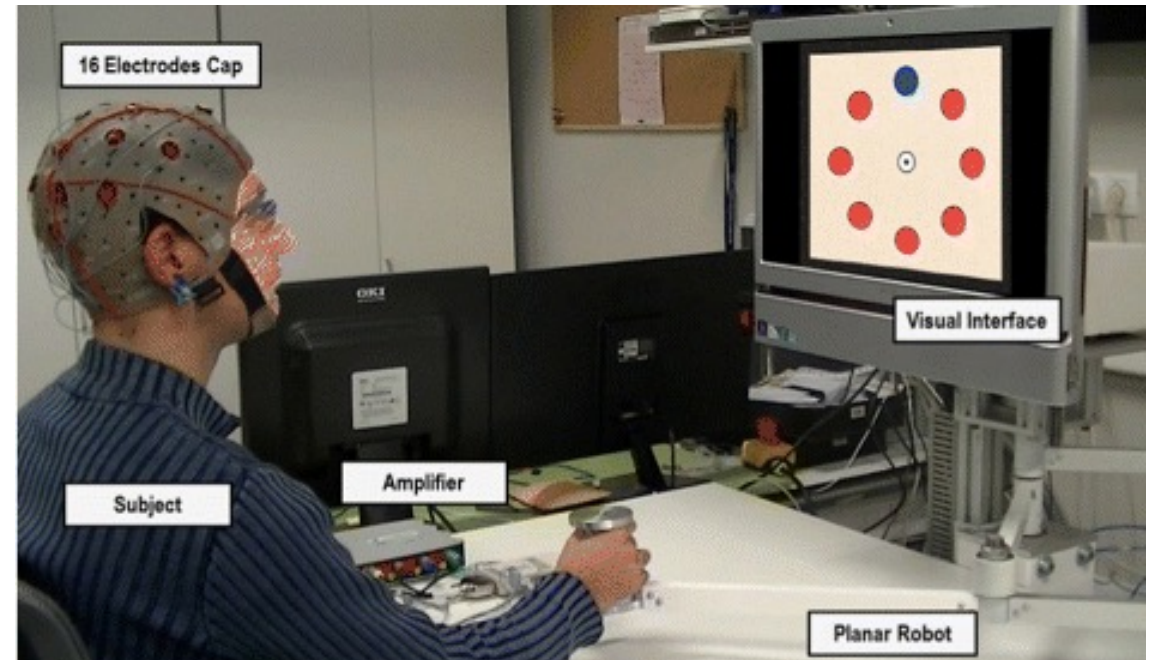
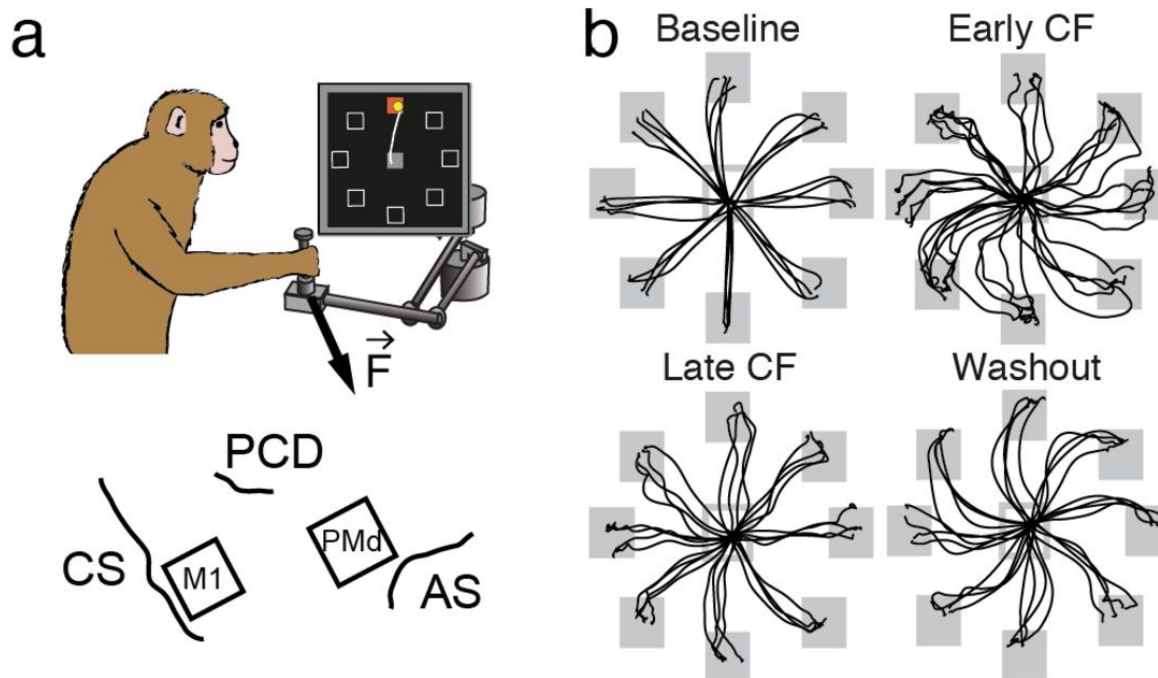
Evarts 1968, *J. Neurophysiology*

Center-out motor task

The center-out task is frequently used in motor control studies.

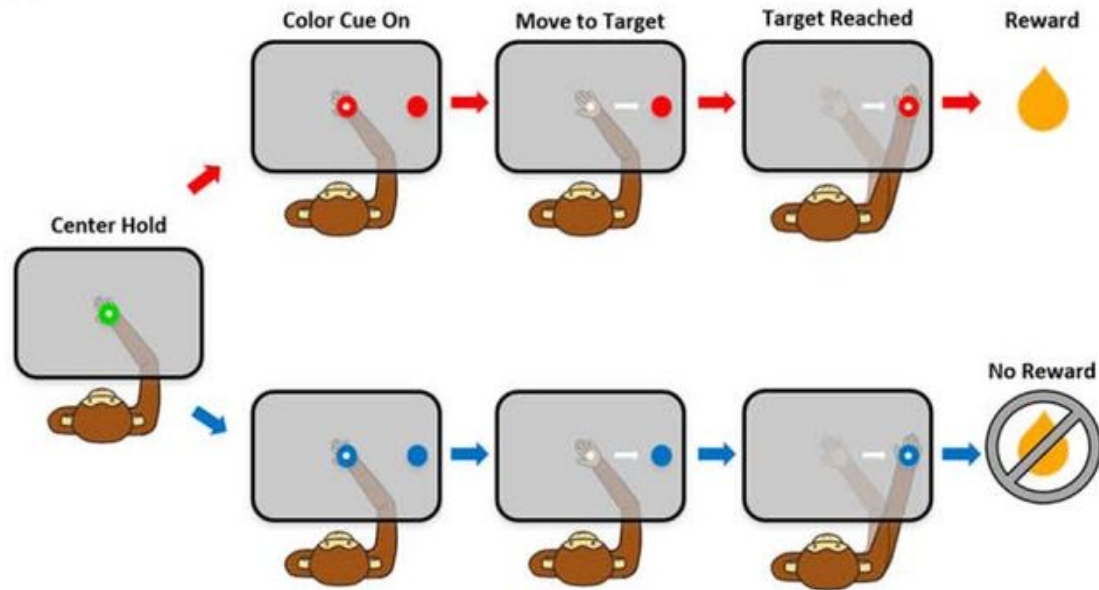
The task is very separable yet constrained.

It can be performed with a limb, a cursor, or a robotic manipulandum.

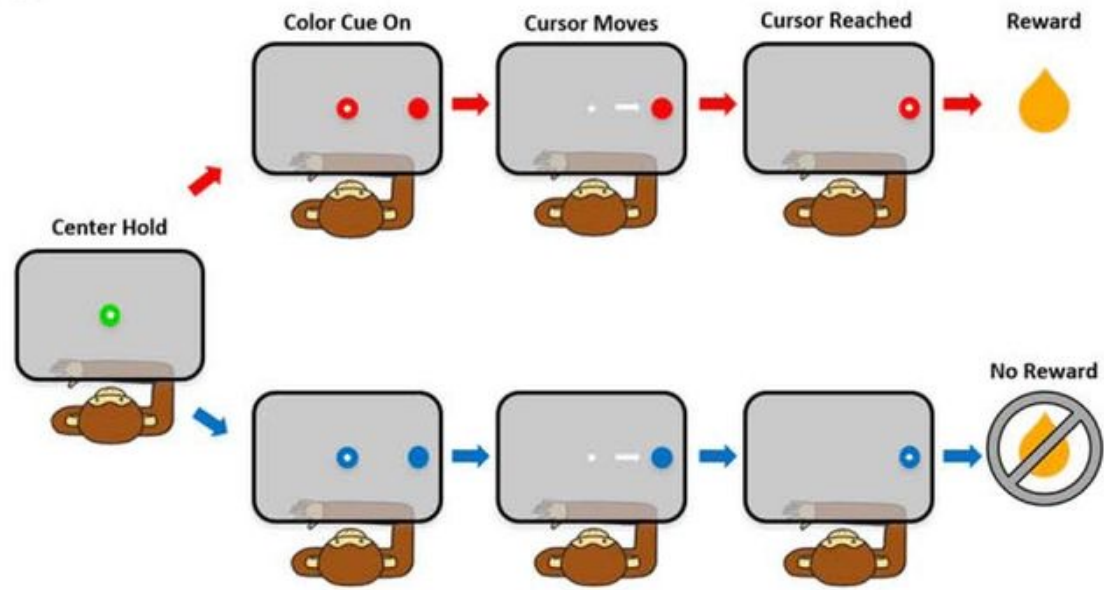


Center-out motor task

A Center-Out Reaching Task (Manual)



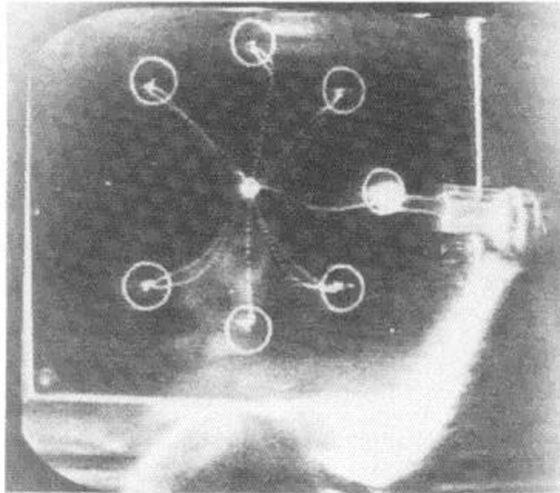
B Center-Out Reaching Task (Observational)



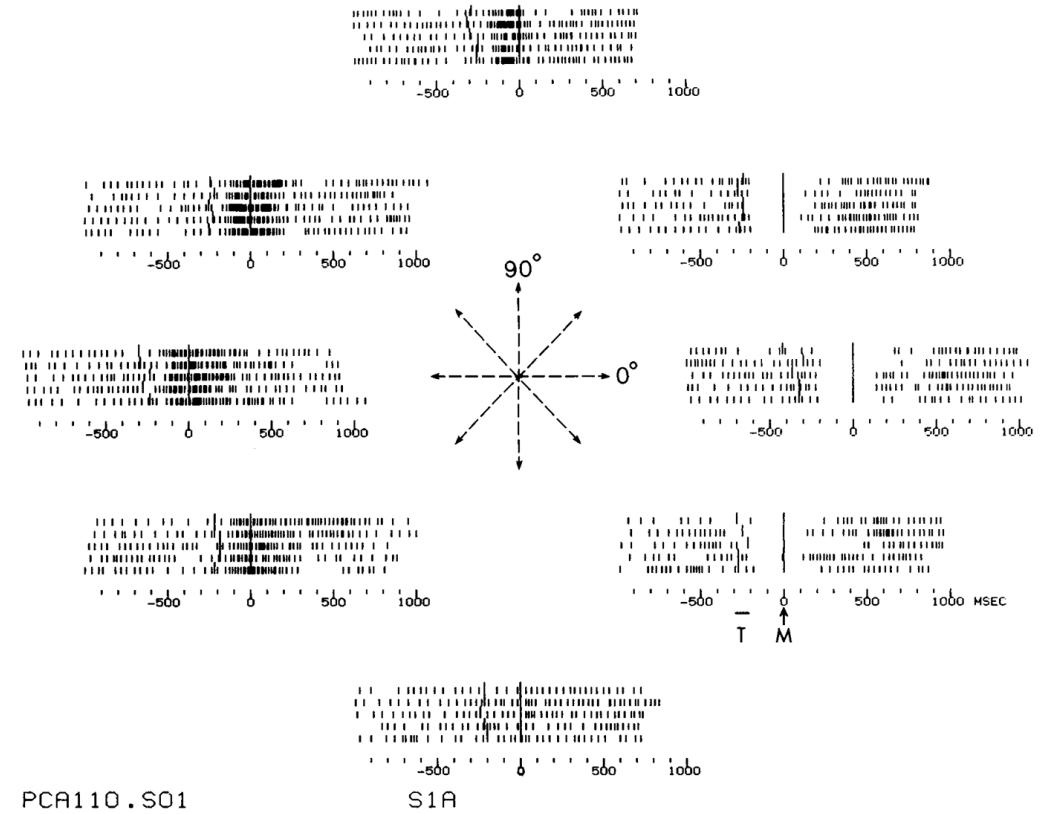
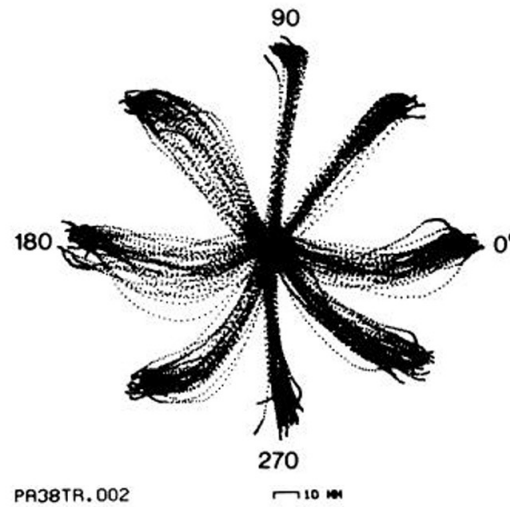
Observational version of the center out task are also used to train neural decoders.

How does motor cortex encode movement?

B.



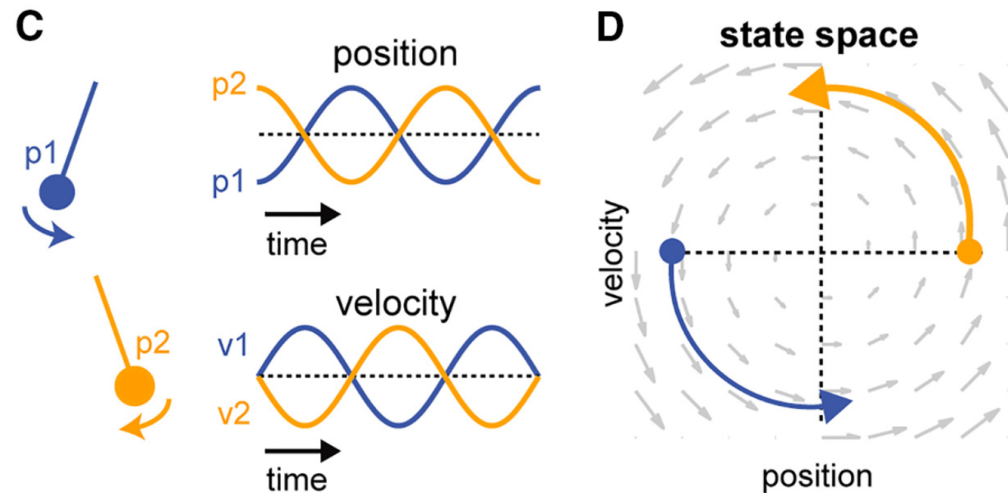
C.



Georgopoulos et al., 1982 *J. Neuroscience*

Review: what are dynamical systems?

“A set of coordinates, often represented as a vector, describing the instantaneous configuration of a dynamical system and that is sufficient to determine the future evolution of that system and its response to inputs.”

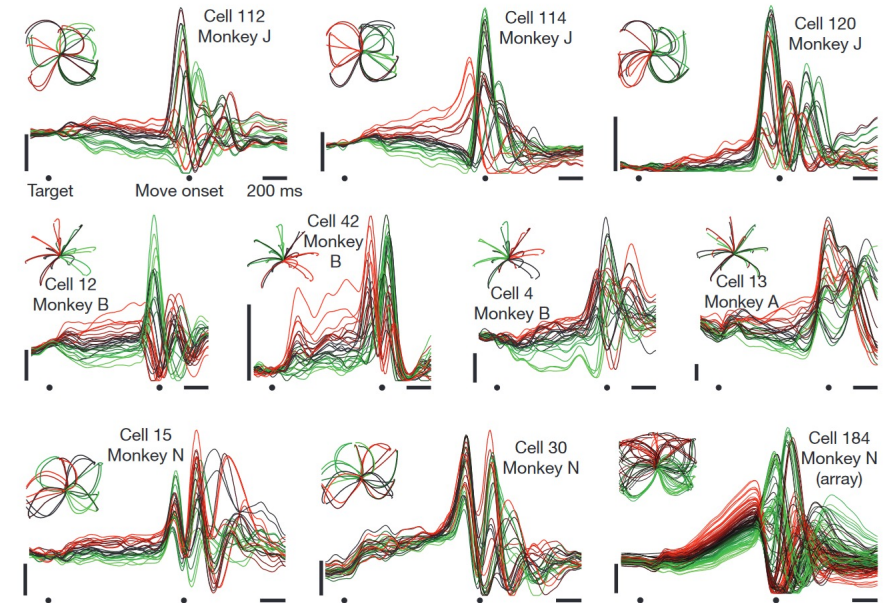
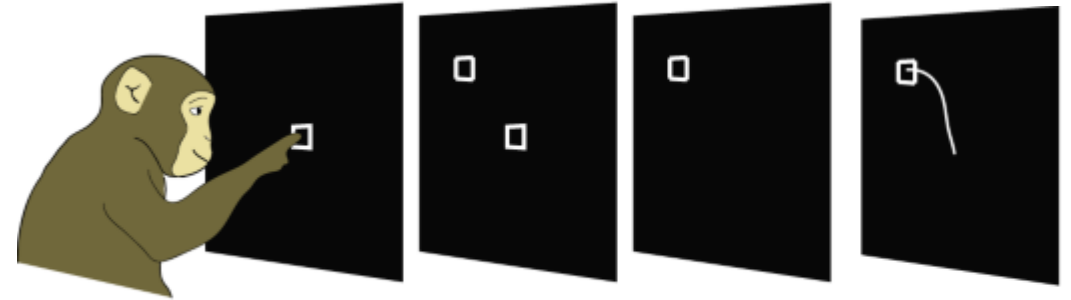


Churchland Shenoy 2013
Ann. reviews in neuroscience

Pandarínath et al., 2018 *J neuroscience*

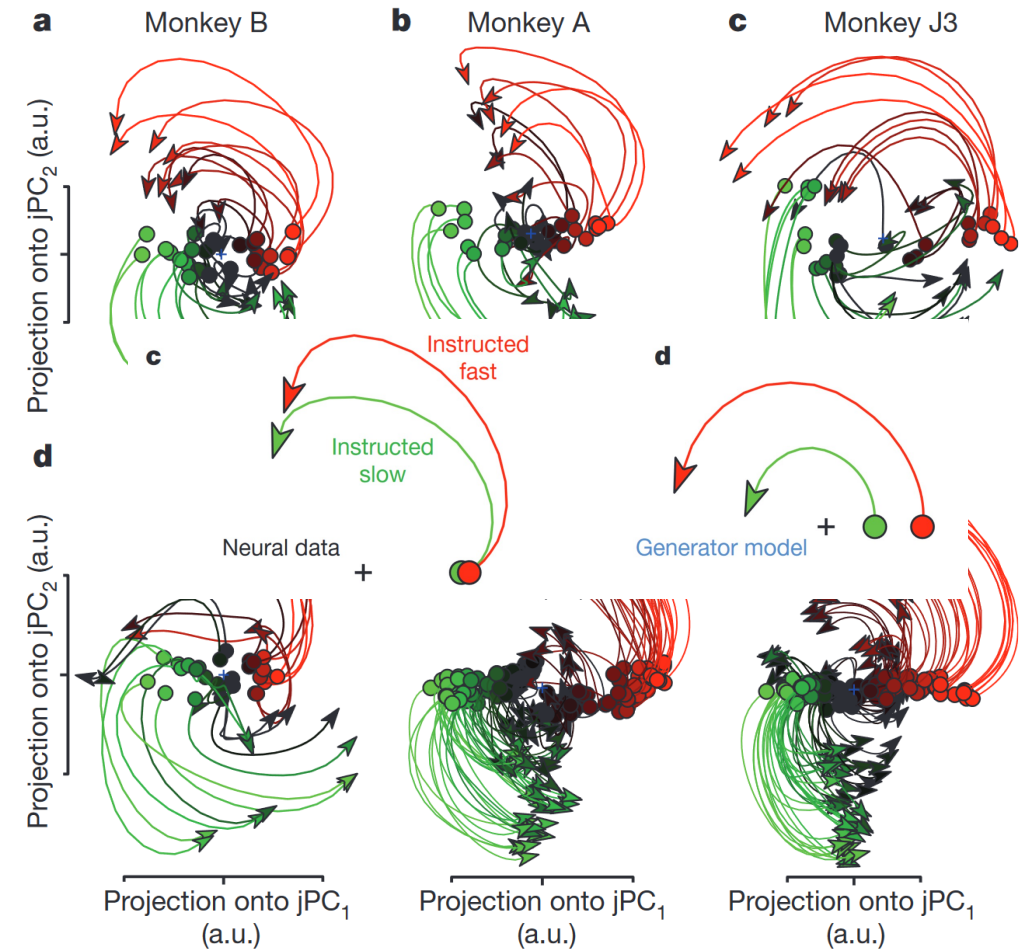
Discovery of rotational dynamics

- Monkeys performed a center out reaching task while neurons were recorded in primary motor cortex.
- Researchers also recorded from leeches swimming and walking monkeys.

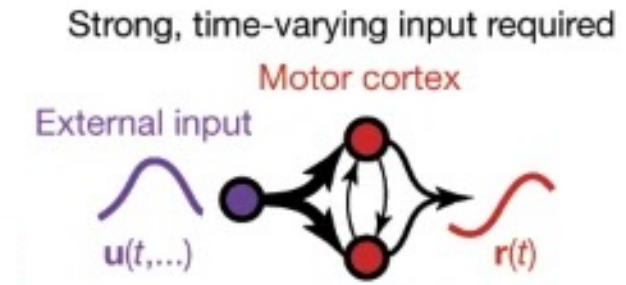
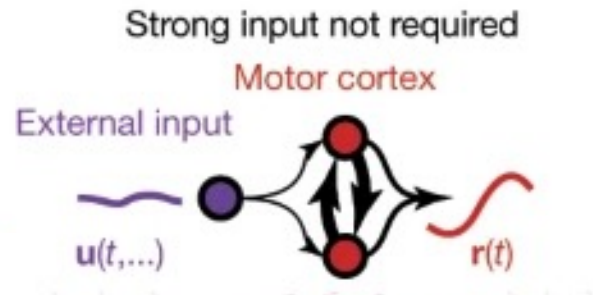


Rotational dynamics

- Low dimensional projections of neural activity during center-out reaching tasks produce highly consistent neural trajectories.
- These cyclical trajectories appear to show organization based on movement kinematics including direction and velocity of movements.
- This finding suggests that motor cortex acts as a dynamical system, with neural activity evolving over time based on local dynamics and external inputs.



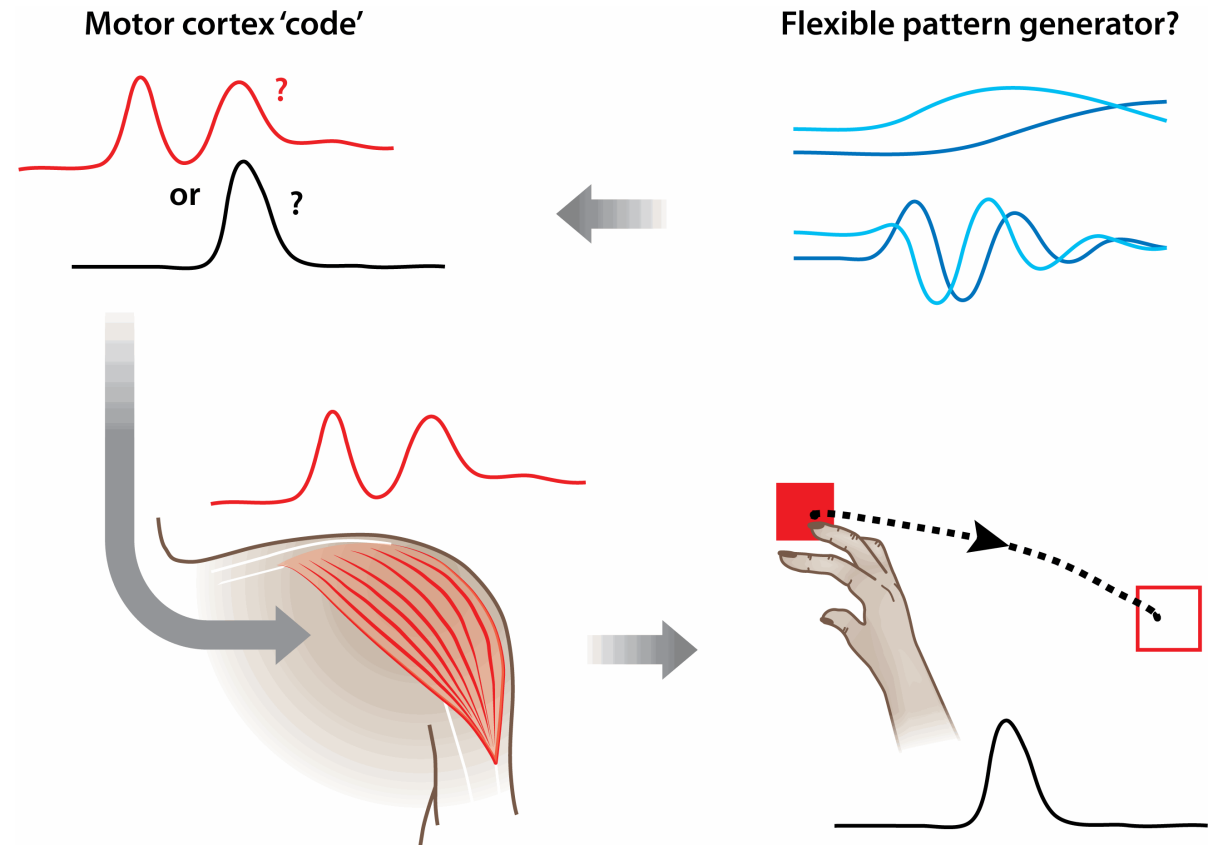
Motor cortex as a dynamical system



There is ongoing research into the extent to which motor cortex is input-driven compared to intrinsically driven.

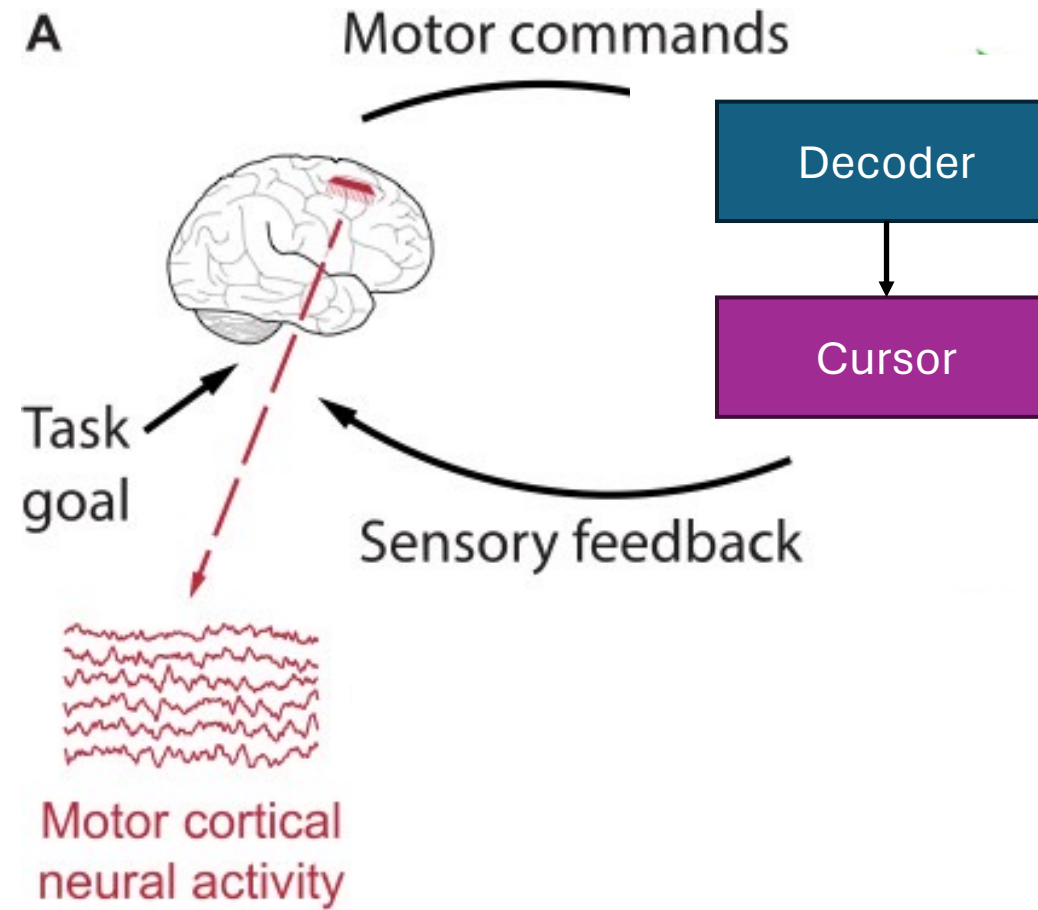
How does motor cortex encode movement?

- How cortical motor commands are transformed into motor movements is still an area of open debate.
- Movement coding could be organized through muscle activations, movement kinematics, muscle synergies or some combination of factors.
- Since the precise transformation from neural cortical activity to movement is still under research, scientists can try to simplify cortical outputs to facilitate research.

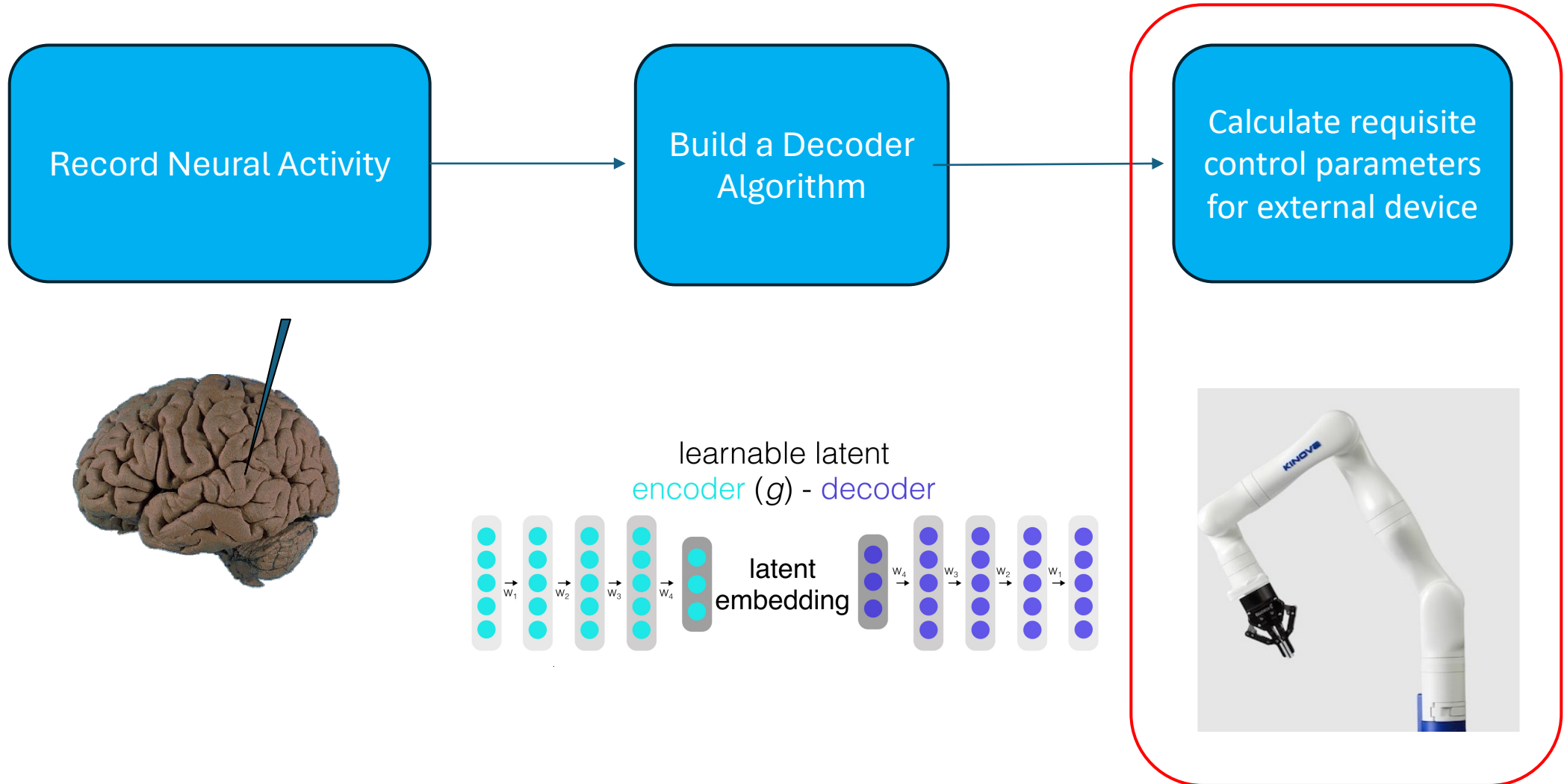


Shenoy KV, et al. 2013.
Annu. Rev. Neurosci. 36:337–59

BMI facilitates causal testing of the effect of neural activity on 'movements'



Simple overview



Can we demonstrate a causal relationship between invariant dynamics in motor cortex and a task-relevant command?



Article

Invariant neural dynamics drive commands to control different movements

Vivek R. Athalye,^{1,7,9,14,*} Preeya Khanna,^{2,7,10,*} Suraj Gowda,⁴ Amy L. Orsborn,^{3,11} Rui M. Costa,^{1,8,12,*} and Jose M. Carmena^{4,5,6,8,13,*}

¹Zuckerman Mind Brain Behavior Institute, Departments of Neuroscience and Neurology, Columbia University, New York, NY 10027, USA

²Department of Neurology, University of California, San Francisco, San Francisco, CA 94158, USA

³Departments of Bioengineering, Electrical and Computer Engineering, University of Washington, Seattle, Seattle, WA 98195, USA

⁴Department of Electrical Engineering and Computer Sciences, University of California, Berkeley, Berkeley, CA 94720, USA

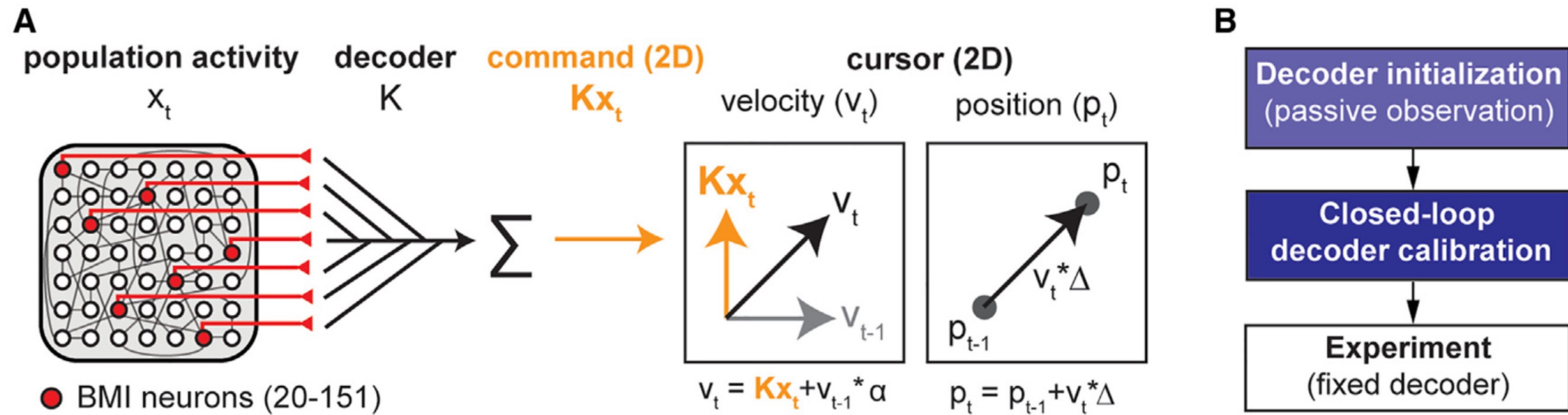
⁵Helen Wills Neuroscience Institute, University of California, Berkeley, Berkeley, CA 94720, USA

⁶UC Berkeley-UCSF Joint Graduate Program in Bioengineering, University of California, Berkeley, Berkeley, CA 94720, USA

⁷These authors contributed equally

⁸Senior author

Experiment setup: training the BMI decoder

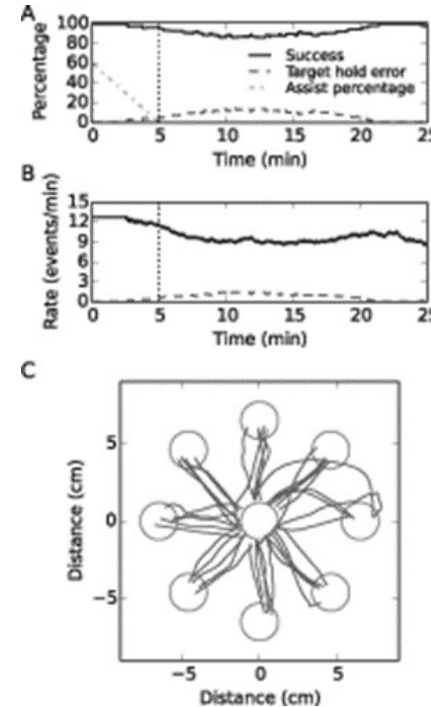
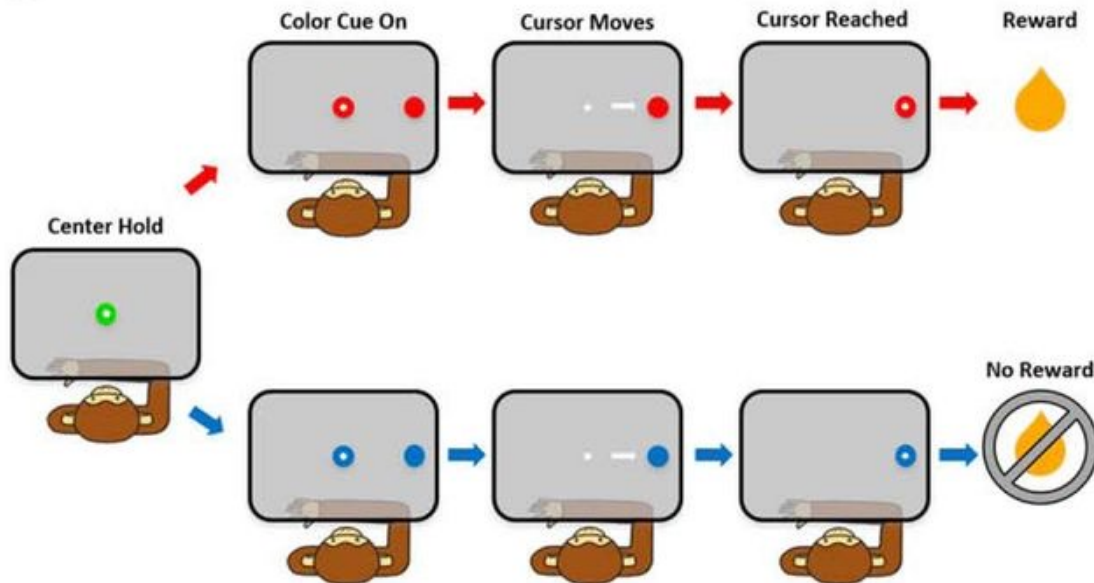


16x8 multi-shank arrays were implanted in the upper limb area of motor cortex.

Monkeys were trained to control a cursor using the implanted array, but without moving their arm.

Experiment setup: training the BMI decoder

B Center-Out Reaching Task (Observational)

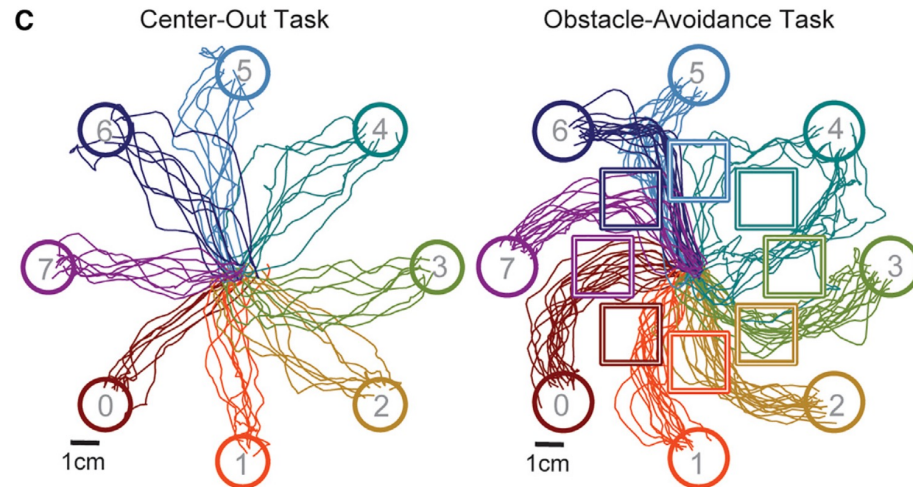


Monkeys first observed the center out cursor task so that decoder units could be identified.

After units were selected, monkeys began training to move the cursor though either a velocity Kalman filter (monkey G) or a velocity Point Process filter (monkey J).

The decoder was refined using “closed-loop decoder adaptation” or CLDA (Danhgi et al., 2014).

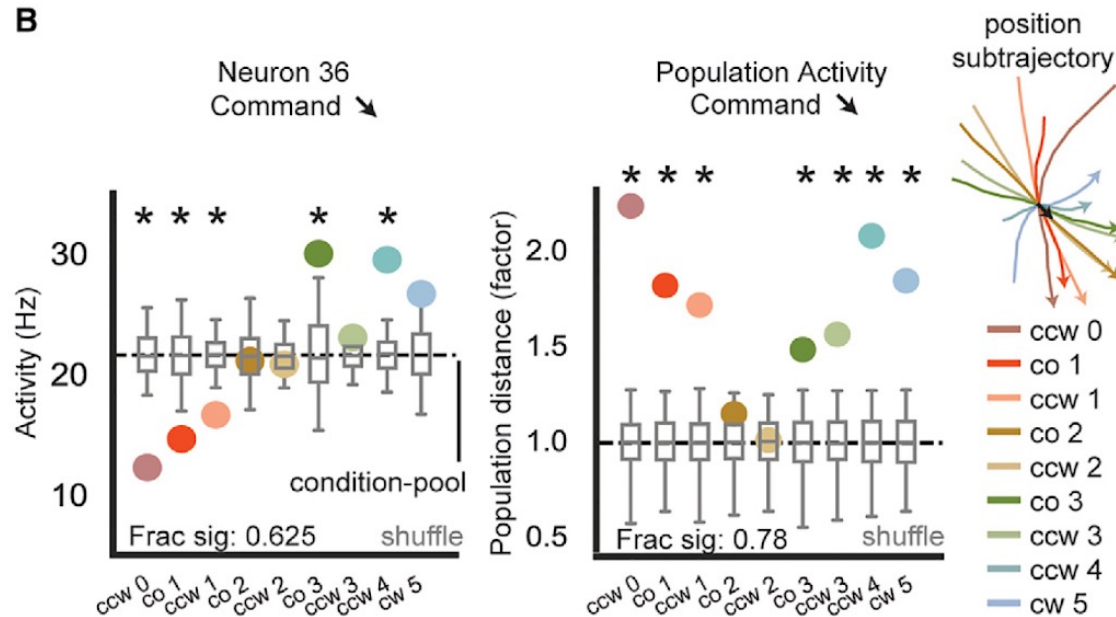
Different neural activity patterns can give rise to the same command.



Animals were trained on two tasks, center-out and obstacle avoidance.

Because the decoder is fixed for both tasks, the motor command produced by the neural activity could be compared across tasks and trials.

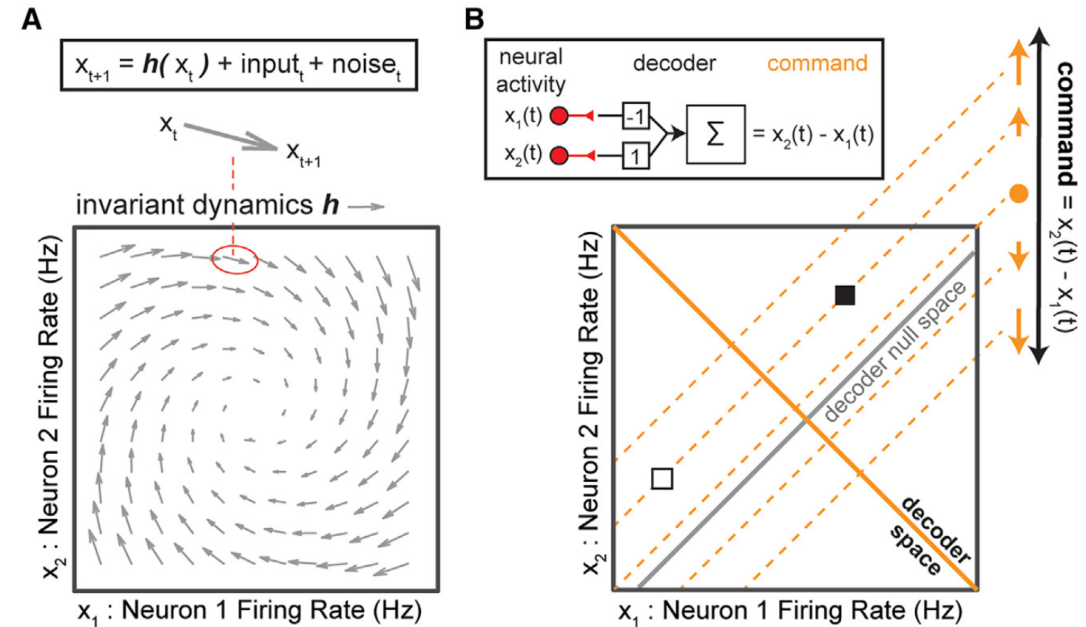
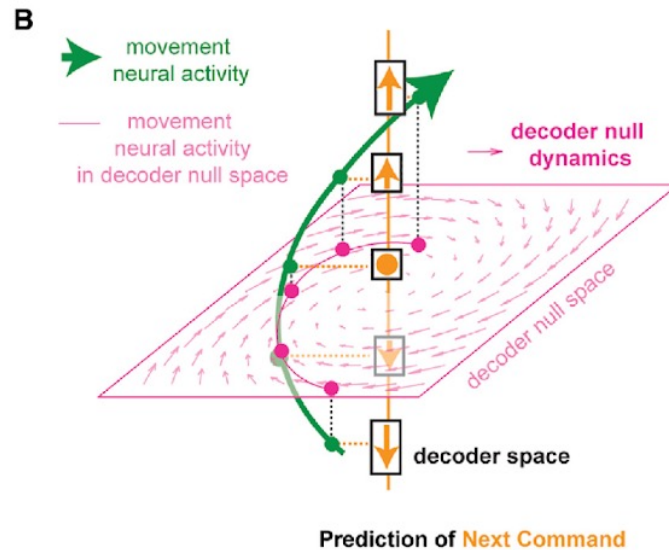
Command-relevant neural activity varies with sub-trajectory



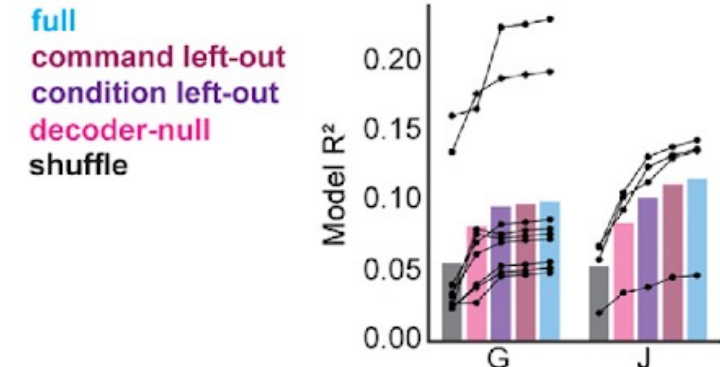
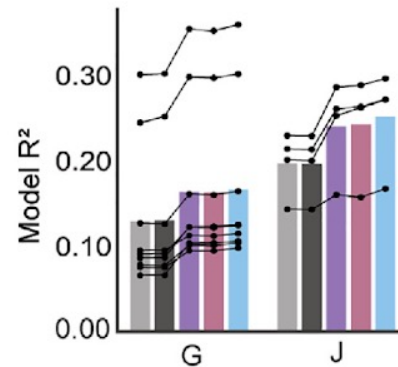
The further a sub-trajectory is from the middle of the trajectory distribution, the more the neural activity producing that command differs from the average activity to produce the command.

Do invariant dynamics are used to control different movements?

Researchers wanted to test whether the invariant dynamics in the neural activity they recorded produced subsequent motor commands in a way that was relevant to the task and condition of each trial.



“Invariant” dynamics predict motor commands in decoder-relevant space



Invariant dynamics can predict the neural activity that produced a motor command, even when task variables are removed.

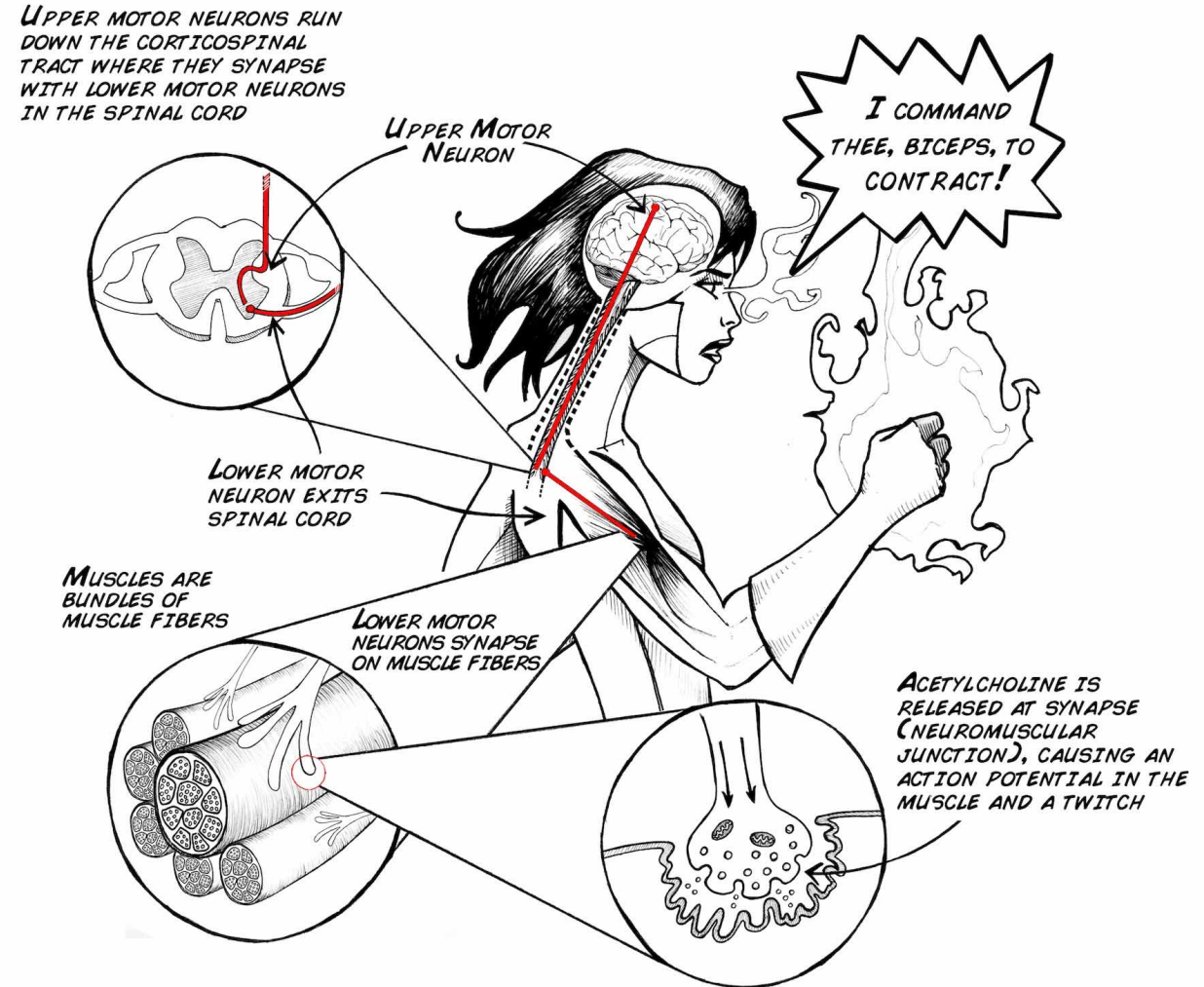
Invariant dynamics can predict the neural activity for the next motor command in a sequence, and the full model outperforms the null-decoder model.

Tomorrow:

Part 1: short (30 min) paper discussion

Part 2:

- EMG recordings in your arm (30 min)
- Making a EMG-based **“B”CI** !



<https://backyardbrains.com/experiments/muscleSpikerBox>

Summary

Overview:

- BCIs, or Brain Computer Interfaces, are systems that facilitate a **direct communication pathway between a brain and an external device**. This technology enables individuals to control devices using only their brain signals.
- **Recording neural activity is the foundation of how BCIs operate**. Specialized algorithms, known as decoders, are then employed to interpret these signals into commands that can control devices or computer systems.
- The importance of (**encoder-**) **decoder** algorithms lies in their ability to translate neural activity into actionable instructions for external devices, making them integral to the functionality of BCIs.

Systems Neuroscience Contributions:

- Instrumental in identifying optimal brain areas for signal recording, understanding neural subtypes, and designing effective sensory feedback within BCIs.
- Insights into neural dynamics, such as the relationship between neural firing and sensory stimuli or motor actions, thereby informing the development of more advanced BCIs.
- Current research in systems neuroscience contributes to BCIs by examining the principles of encoding sensory information into neural activity and decoding it back into the brain.

Current Technologies in BCIs:

- **Microelectrode arrays** are a key technology in BCIs, allowing for the stable recording of neural activity over extended periods. These arrays can be implanted and have been used in both research settings and, to a lesser extent, in clinical applications to assist individuals with paralysis.
- **Two-photon holographic optogenetics** represents a cutting-edge approach in BCI technology. It enables precise manipulation and recording of neural activity using light (calcium imaging and optogenetics).
- Technological advancements in BCI include increased recording stability and longevity, more biocompatible materials for implants, and higher throughput in signal recording. These improvements are crucial for the reliability and user-friendliness of BCIs, ultimately enhancing their applicability and integration into various aspects of life and healthcare. Ethics are also deeply important to consider.