

# Neuroscience view on machine learning

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TOOPLOOX AI

# Cognitive science

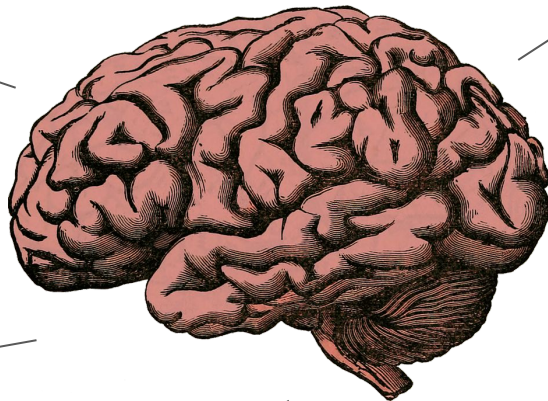
Psychology

Neuroscience

Philosophy

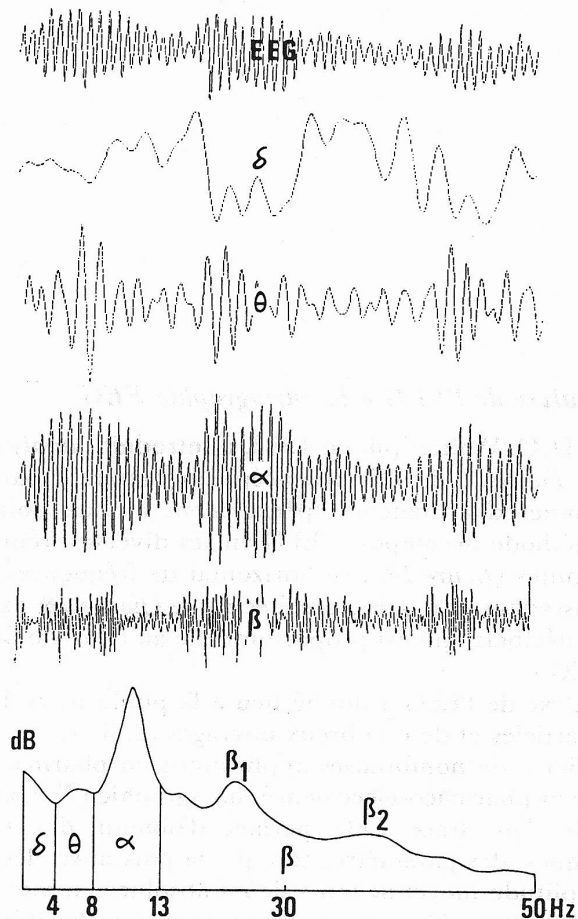
Artificial Intelligence

Linguistics



## Signals of interest

- Behavioral (how fast did this person press a button?)
- Physiological heart rate, pupil dilation, EEG etc.
- **NEW: digital** (indirect collection of data from digital resources, e.g. Twitter, Reddit, Netflix, Google Analytics etc. )



# Artificial Intelligence & Neuro/Cognitive Science

How do they go together?

Cognitive science → Artificial Intelligence

- **Cognitive modelling** - create formal models of human's cognition and test their validity
- Influence artificial intelligence with human-like cognitive structures (attention, memory, visual perception)
- Neural networks resembling real neural structures (**spiking neural networks**)
- **Intelligent agents** that mimic human's cognition

Artificial Intelligence → Cognitive Science

- **Computational neuroscience** - use of advanced computational tools in order to understand how brain works
- Use large datasets to investigate cognitive phenomena (e.g. depression prediction, intelligence prediction, **affective computing**)

# Examples

1. **What can you do with your brain/neural signal?**
2. **How can artificial neural network give you knowledge about cognition?**

The background features several thin, curved lines in shades of blue, red, and grey, creating a modern, abstract aesthetic.

# **BCI – Brain-Computer Interface**

# BCI = Brain - Computer Interface

- Communication pathway between brain and computer
- Enables two-way communication (brain -> computer / computer -> brain)

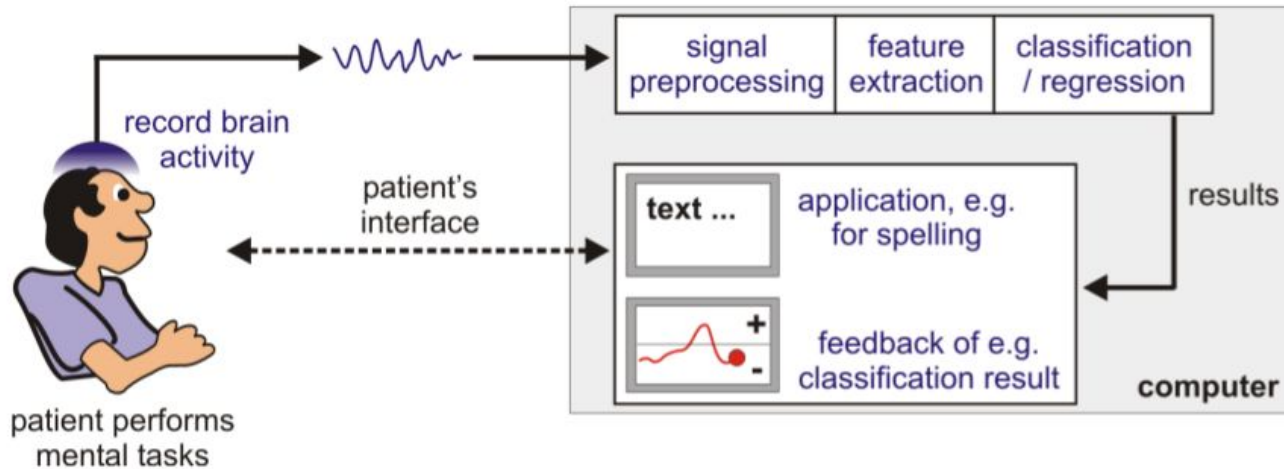
Types of BCIs:

Non-Invasive

Semi-Invasive

Invasive

## BCI - scheme of work



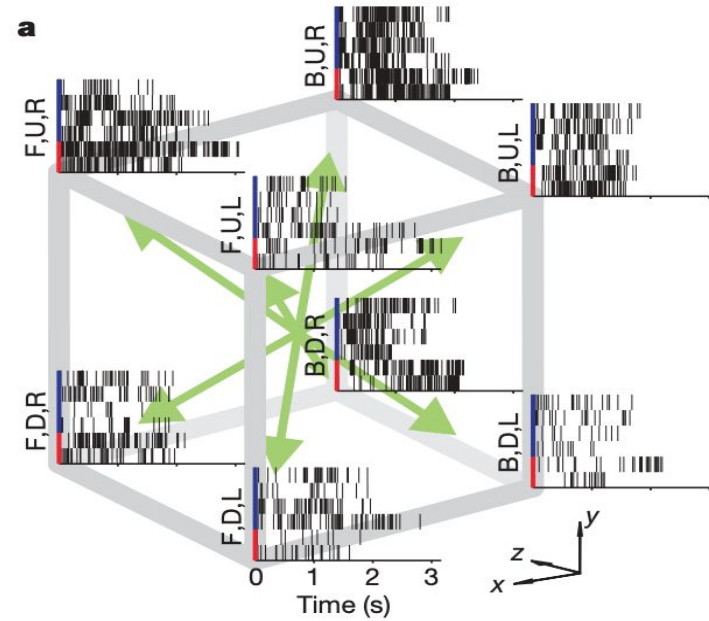


## Do BCI + ML go together? Monkey robotic arm case



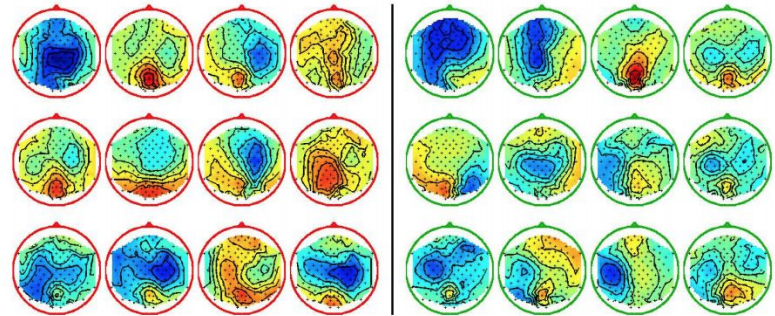
- First, monkeys learned to perform this task with a joystick
- Monkeys trained 10 days - 2 months on a variety of intermediate steps to achieve high performance
- Algorithm used: **PVA: Population Vector Algorithm**
  1. Calculate instantaneous firing rate for each unit
  2. Smooth the response
  3. Population vector = vector sum of preferred directions weighted by the normalized firing rates

## Do BCI + ML go together? Monkey robotic arm case

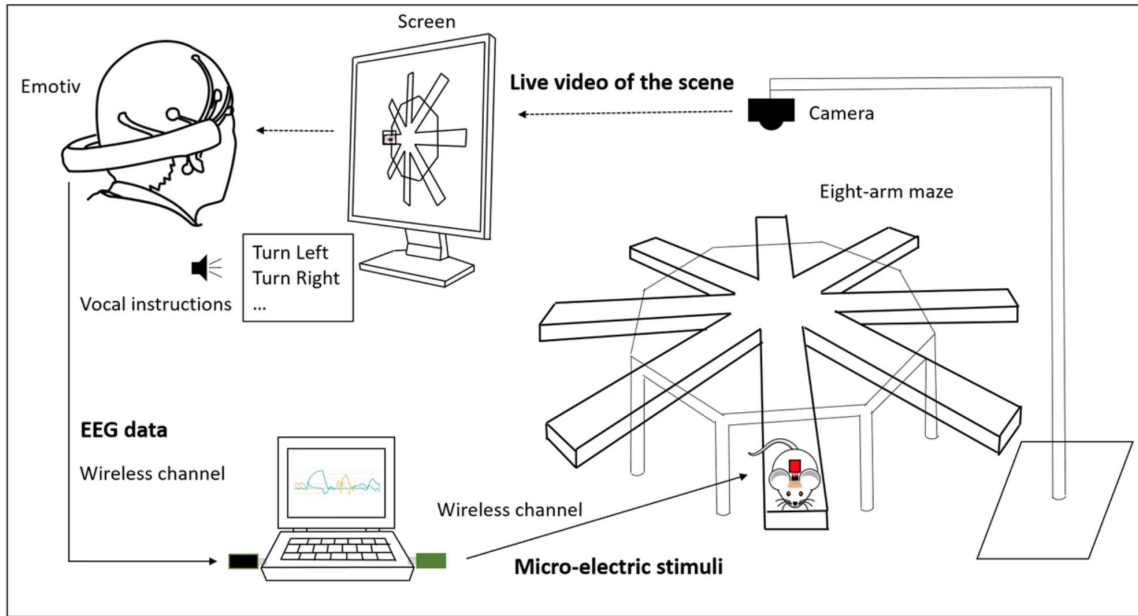


## BCI + ML - challenges

- Great intra-subject variability
- Session-to-session variability
- Trial-to-trial variability
- Ecological validity
- Non-stationary signal



## BBI - rats cyborgs



### Procedure:

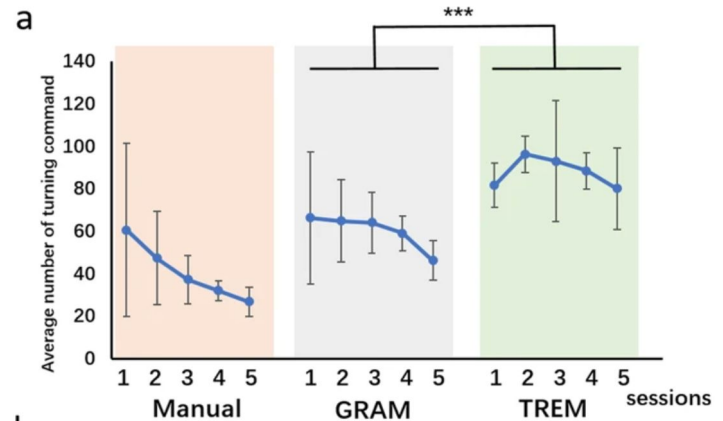
1. Human - trained to express the motor intent for Left/Right movement (blinking signal for forward)
2. Rat - chipped with electrode for virtual reward stimulation; trained to correlate reward stimulation with locomotion behaviors

## BBI - rats cyborgs



Two models for generating Left/Right instructions:

- TREM (instructions generated when passing a threshold)
- GRAM (instruction generated when the gradient value between two possible results exceeded threshold)



## Other cool BCI examples

- [Brain music](#) / [Other brain music](#)
- [Gameplay with no hands involvement](#)
- [Drone control](#)
- [F1 car controlled by quadriplegic](#)

The background features several thin, curved lines in shades of light gray and blue, creating a sense of movement and depth. A prominent blue line curves from the top right towards the center, while a red line follows a similar path slightly below it. Other lighter gray lines are scattered across the frame, some intersecting the more vibrant ones.

# Seeing with a neural network



## Decoding brain activity into interpretable images

Scientific problem: given the brain activity pattern from a fMRI (e.g.), how can we decode it into an understandable percept?

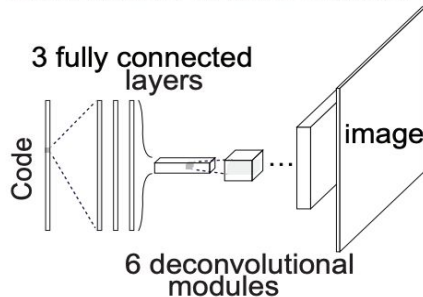
Sci-fi version: can we read people's minds?





# Decoding brain activity into interpretable images

## A Generative neural network



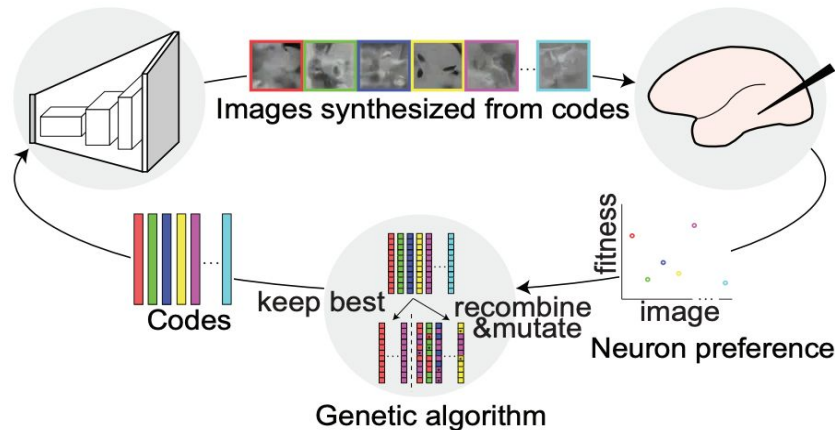
## B Starting images



## D Overall schematic

Generative neural network

Neuronal Recording



## C

+ Fixation point  
Receptive field

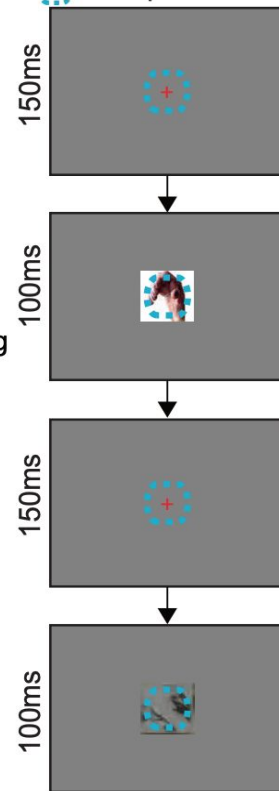
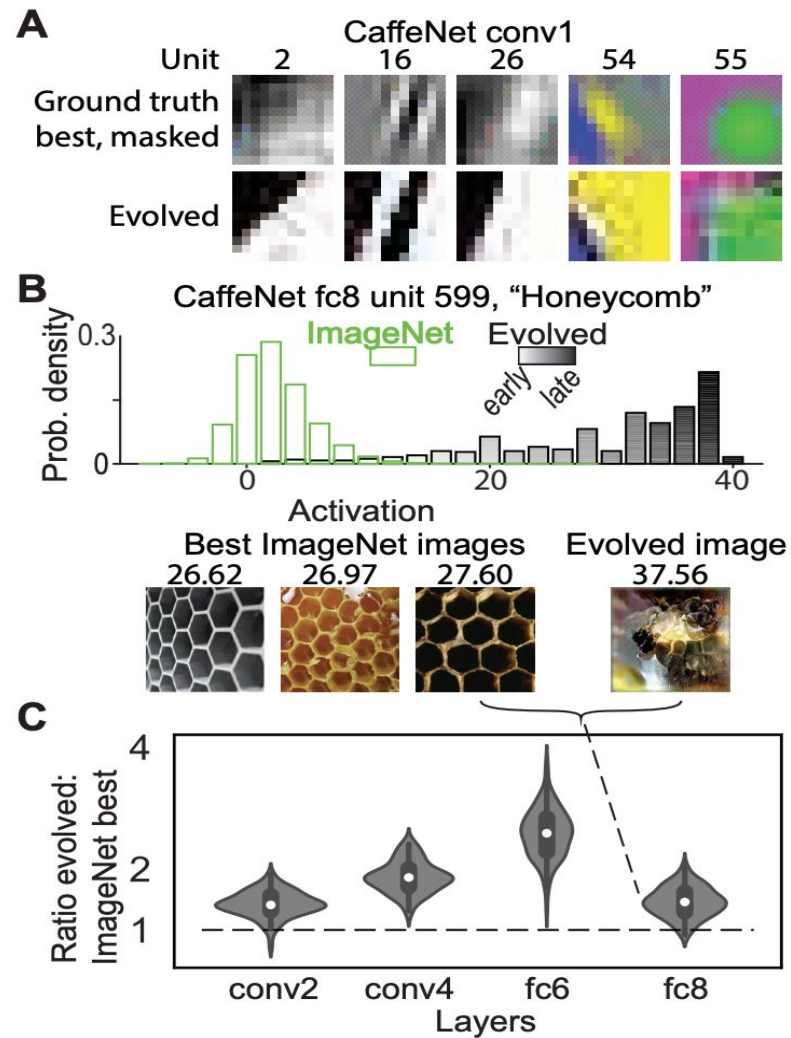


Figure 1. **Synthesis of optimum images via neuron-guided evolution.** (A) Generative network. Architecture of the pre-trained deep generative network (Dosovitskiy & Brox, 2016). The network comprised three fully connected layers and six deconvolutional modules. (B) The initial synthetic images were random achromatic Simoncelli and Portilla textures. (C) Behavioral task. Animals fixated within a  $2.0^\circ$ -diameter window while images were presented for 100 ms followed by a 100 to 200 ms blank period. Red cross: fixation point; dashed line, population RF. (D) Experimental flow. Image codes were forwarded through the deep generative network to synthesize images presented to the monkey.

# Decoding brain activity into interpretable images



# Decoding brain activity into interpretable images

A

Evolved images

Top natural images

Ri-23 (PIT)



Gu-21 (PIT)



Ge-7 (CIT)



Y1-14 (CIT)



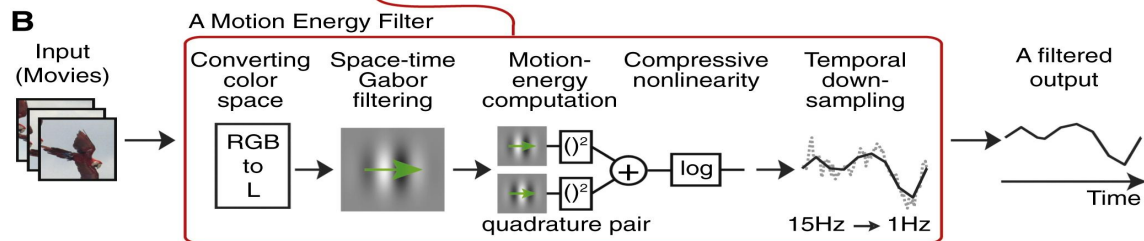
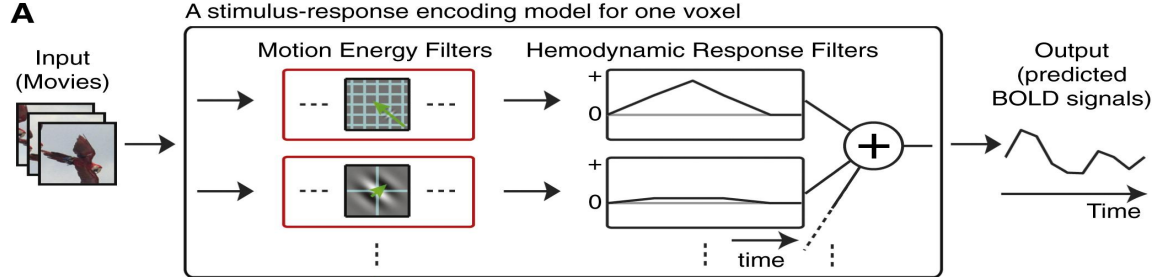
# Decoding brain activity into interpretable images

Link to the video:

<https://ars.els-cdn.com/content/image/1-s2.0-S0960982211009377-mmc2.mpg> (download begins)



# Decoding brain activity into interpretable images



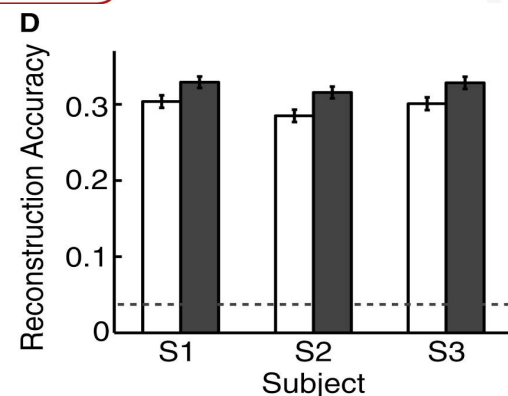
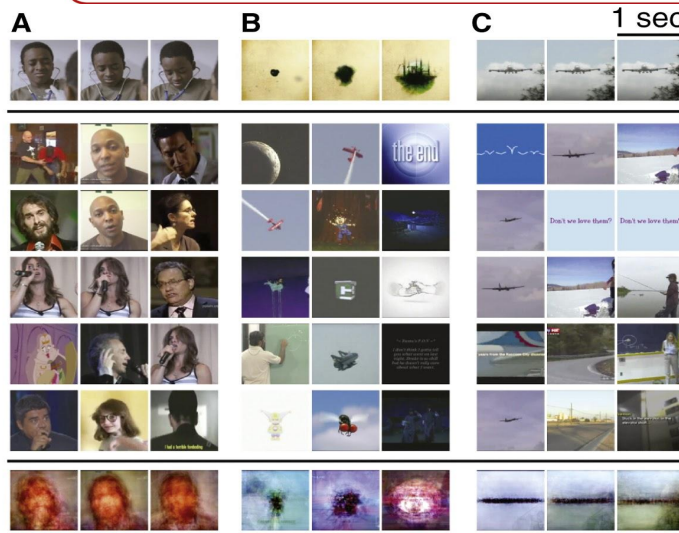
Presented movies

Highest posterior movies (MAP)

3rd highest

5th highest

Reconstructed movies (AHP)



□ MAP reconstruction  
 ■ AHP reconstruction  
 --- Chance level ( $p=0.01$ )

## Don't know where to start?

- Recently released Google Dataset Search:  
<https://datasetsearch.research.google.com/>
- And maybe my [blogpost](#) focused on the brain data repositories
- [Brainhack Warsaw 2020](#)
- Useful concepts: emotion recognition, visual perception, neuroprosthetics, psychophysiology, biohacking, sensory models

# References

1. [Brain computer interfacing: Applications and challenges](#)
2. [Human Mind Control of Rat Cyborg's Continuous Locomotion with Wireless Brain-to-Brain Interface](#)
3. [Cortical control of a prosthetic arm for self-feeding](#)
4. [Evolving super stimuli for real neurons using deep generative networks](#)
5. [End-to-End Deep Image Reconstruction From Human Brain Activity](#)
6. [Reconstructing Visual Experiences from Brain Activity Evoked by Natural Movies](#)

**If you want to chat more  
about brainz&computerz,  
I will be very happy to  
answer:**

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