





Artificial Intelligence and Neurotechnologies for Human Augmentation



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Frontiers of Artificial Intelligence and Machine Learning, China, 14-16.04.2023

Toruń



Toruń



Nicolaus Copernicus, born in 1473 in Toruń (550 years ago). Studied in Krakow, Bologna, Padova and Ferrara.

Al ⇔ Brain

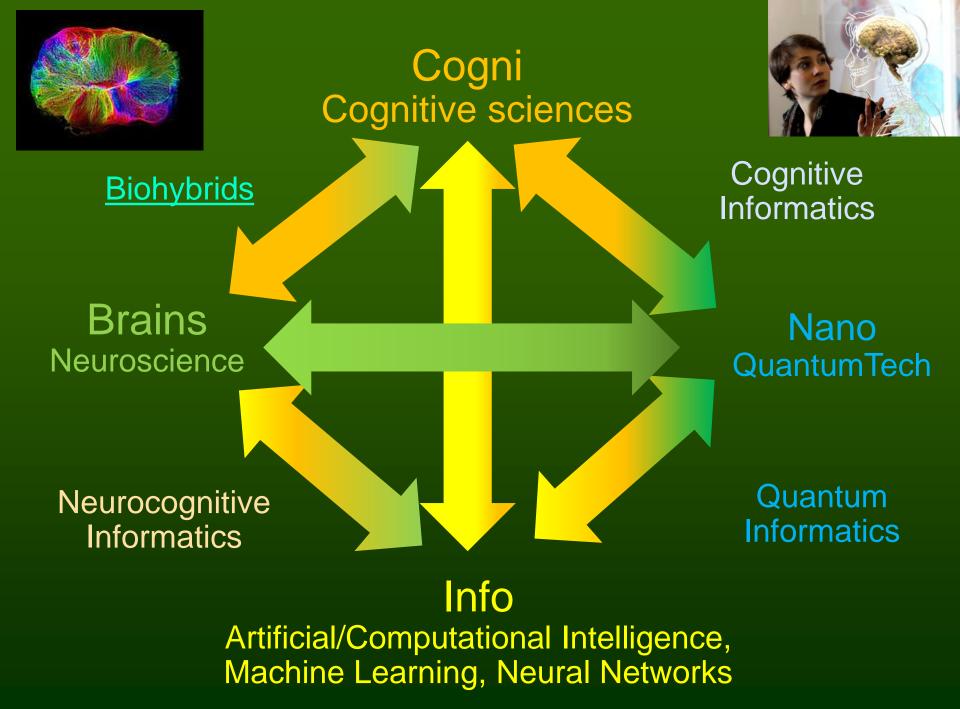


AI

- Large Language Models.
- Generative Pretrained Transformers (GPT).
- Attention mechanism and the brain.

Brains

- Human enhancement with neurocognitive technologies.
- Closed loop biofeedback.
- Source localization, spectral fingerprinting, recurrence.
- Unrealistic expectations.



Artificial Intelligence

AGI & BICA

From an engineer's perspective, to understand the brain is to build a working model that exhibits the same functions. Needed: spatial models of phenomena, actions and their causes, real world imagery.

AGI = Artificial General Intelligence, learn many different things.

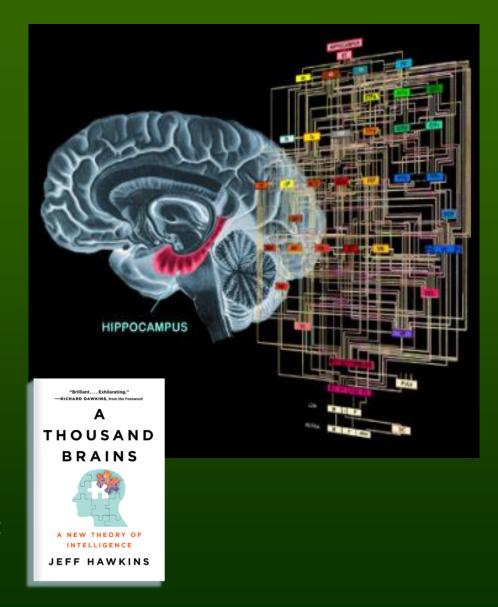
BICA (Brain-Inspired Cognitive Architecture) brain-like intelligence.

Duch, Oentaryo, Pasquier,

<u>Cognitive architectures:</u> where do we go from here?

"We'll never have true AI without first understanding the brain"

Jeff Hawkins (2020).



Artificial General Intelligence (AGI), Memphis 2008



2022: <u>DeepMind Gato</u> is a relatively small model, with 1.2 billion parameters. Multi-modal, multi-task, multi-embodiment, learned simultaneously over 600 tasks, games to controlling robots. Small working memory capacity.

Towards Human-like Intelligence

IEEE Computational Intelligence Society Task Force (Mandziuk, Duch, M. Woźniak),

Towards Human-like Intelligence



IEEE SSCI CIHLI 2022 Symposium on Computational Intelligence for Human-like Intelligence, Singapore. In 2023 it will be in Mexico City.

AGI: conference, Journal of Artificial General Intelligence comments on Cognitive Architectures and Autonomy: A Comparative Review (eds. Tan, Franklin, Duch).

BICA: Annual International Conf. on Biologically Inspired Cognitive Architectures, 13th Annual Meeting of the BICA Society, Guadalajara, Mexico 2023.

Brain-Mind Institute Schools, International Conference on Brain-Mind (ICBM) and Brain-Mind Magazine (Juyang Weng, Michigan SU).

Neuroscience ⇔ Al

Hassabis, D., Kumaran, D., Summerfield, C., Botvinick, M. (2017). Neuroscience-Inspired Artificial Intelligence. *Neuron*, *95*(2), 245

Affiliations: Google DeepMind, Gatsby, ICN, UCL, Oxford.

Attention, awareness models, consciousness, complementary learning systems, various types of memory, reinforcement learning are used in machine learning.

Key concepts from RL inform neuroscience and ML techniques are basic tools for analysis and interpretation of brain neuroimaging data. Ex:

CNN ⇔ interpret neural representations in high-level ventral visual stream of humans and monkeys, finding evidence for deep supervised networks.

LSTM architecture provides key insights for development of working memory, gating-based maintenance of task-relevant information in the prefrontal cortex.

All this will help in development of neurocognitive technologies.

Language algorithms



Language models: relation of words in complex network structures. In 2018, to gain a general-purpose "language understanding", Google created BERT, model pre-trained on a very large text corpus.

- <u>Bidirectional Encoder</u> Representations from Transformers (BERT). <u>Transformer</u>-based <u>machine learning</u> technique for (NLP) pre-training.
- English-language BERT: two networks, smaller 110M parameters, larger model with 340M parameters in 24-layers; trained on the BooksCorpus with 800M words, and Wikipedia with 2,500M words. In 2019 BERT worked already in 70 languages.
- BERT model was then fine-tuned for specific NLP tasks such as question answering or semantic information retrieval. Many smaller pre-trained open software models were published in GitHub repository.
- The network learns to predict masked words (images, signals):
 Input: the man went to the [MASK1]. He bought a [MASK2] of milk.
 Labels: [MASK1] = store; [MASK2] = gallon.

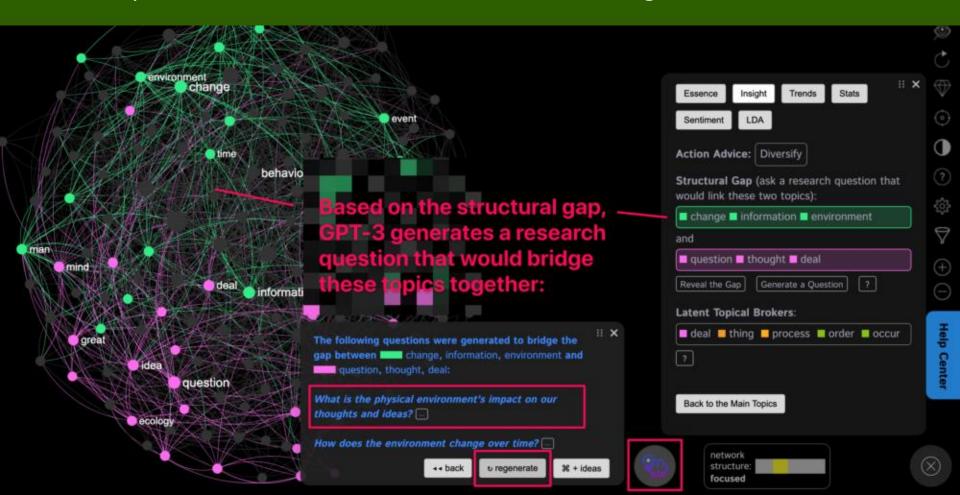
Vision-language models

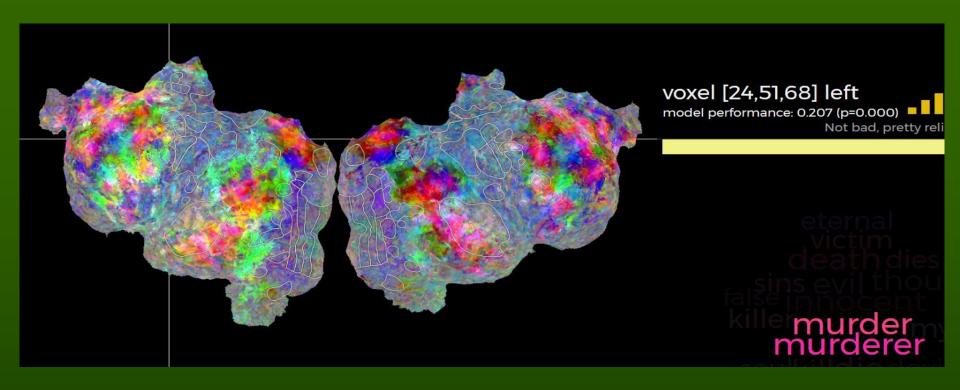
Vision-Language Pre-Trained Models (VL-PTMs), convergence of language, vision, and multimodal pretraining => general-purpose foundation models can handle be easily adapted to multiple diverse tasks with zero-shot learning.



Large language generative models

In the space of trillion parameters many tokens may be linked – such entities as verbal concepts, images, internal body states in robots in Palm-E, BEIT, Gato ... Basic mechanisms involve masking, attention, positional encoding, reflection. Still why these models work so well is not clear. Priming subnetworks?





Whole fMRI activity map for the word "murder" shown on the flattened cortex.

Each word activates a whole map of activity in the brain, depending on sensory features, motor actions and affective components associated with this word.

Why such activity patterns arise? Brain subnetworks connect active areas.

http://gallantlab.org/huth2016/ and short movie intro.

Can one do something like that with EEG or MEG? Prompts invoke specific activation in LLMs.

Creativity: a simple model

My computational creativity project (2005):

Model inspired by brain process involved in creating new names.

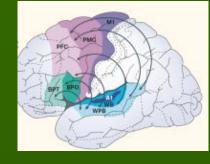
- make a simplest test for creative thinking invent novel names;
- create interesting new names for products, capturing their characteristics;
- understand newly invented words that are not in the dictionary.

Assumption: a set of keywords (prompts) primes the trained cortex subnetwork.

Prompts: ordered strings of phonemes that activate semantic reps of words. Spreading activation => context priming + inhibition in the winner-takes-all process leaves only a few semantically related concepts (word meaning).

Creativity = imagination (activation of pre-trained networks) + filtering (competition, associations to existing states)

Imagination: many transient patterns of activity arise in parallel, activating both words and non-words, guided by the strength of connections **Filtering**: associations, emotions, phonological and semantic density. Is this what happens in large language models?

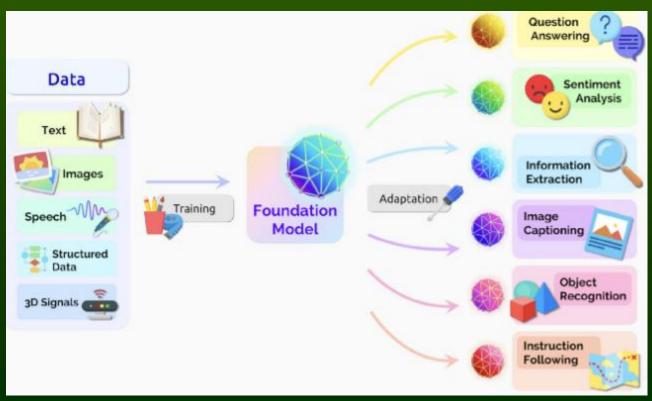


Multimodal models

Multimodal learning – different types of modalities with different statistical properties, embedded in **the same model**.

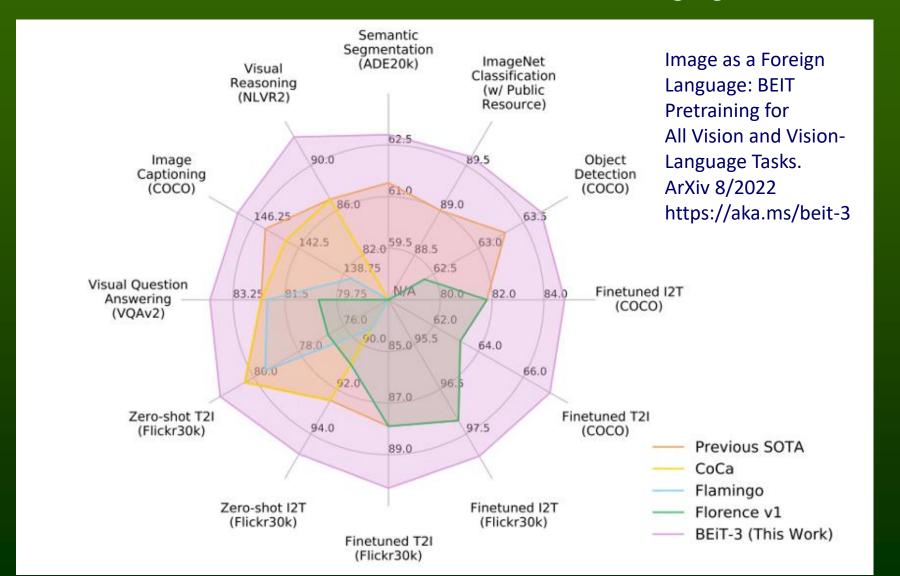
- Multimodal Affective Computing (MAC), sentiment analysis.
- Natural Language for Visual Reasoning (NLVR).
- Multimodal Machine Translation (MMT).
- Visual Retrieval (VR) and Vision-Language Navigation (VLN).

Image: Center for
Research on
Foundation Models
(CRFM), Stanford
Institute for HumanCentered Artificial
Intelligence (HAI)



Vision-language models

MS BEIT-3 (BERT Pretraining of Image Transformers), a general-purpose state-of-the-art multimodal foundation model for vision-language tasks.



Brains

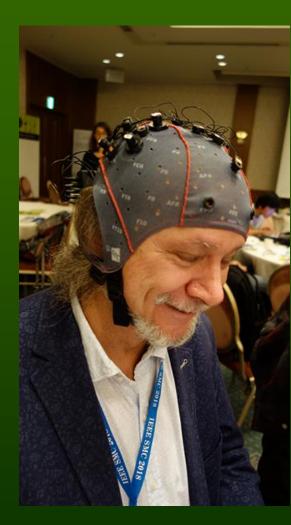
On the threshold of a dream ...

Final goal: optimize brain processes!

Although whole brain is always active we are far from achieving full human potential.

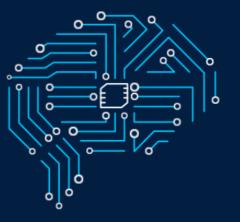
Repair damaged brains, increase efficiency of healthy brains! First we need to understand brain processes:

- 1. Find fingerprints of specific activity of brain structures using new neurotechnologies.
- Create models of cognitive architectures that help to understand information processing in the brain.
- 3. Create new diagnostic and therapeutic procedures.
- 4. Use neurofeedback based on decoding and changes in connectivity to stimulate the brain.
- 5. Stimulate neuroplasticity by monitoring brain activity and directly stimulating it (TMS, DCS, EM).



G-tec wireless NIRS/EEG on my head.



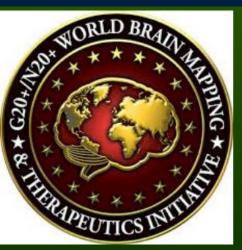


Advance Neurotechnologies

Accelerate the development and application of new neurotechnologies.



Support multi-disciplinary teams and stimulate research to rapidly enhance current neuroscience technologies and catalyze innovative scientific breakthroughs.

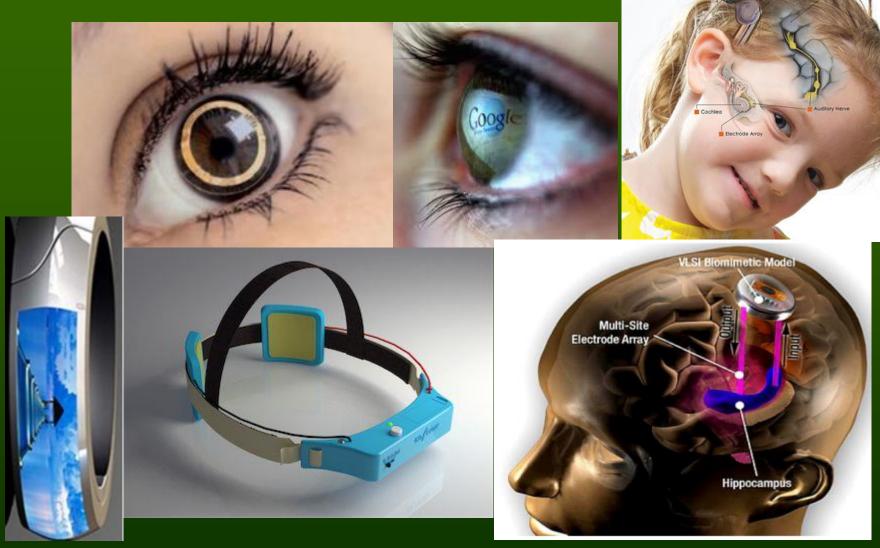






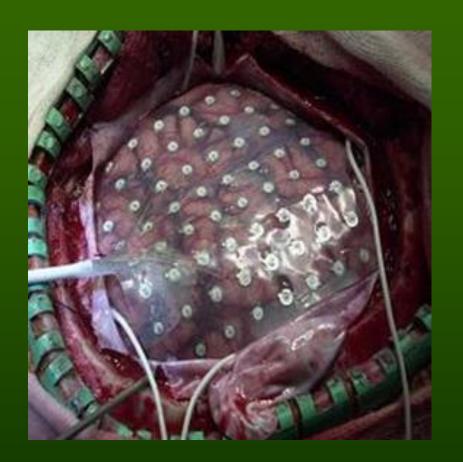
Human Brain Project, EU Flagship (2013), and Obama BRAIN Initiative (2013): BRAIN=Brain Research through Advancing Innovative Neurotechnologies.

Amplification



Improvement of our senses: sight, hearing, touch, memory, attention ... Improving brains by adding new senses (Eagleman, Livewired 2020).

Brain computer interfaces

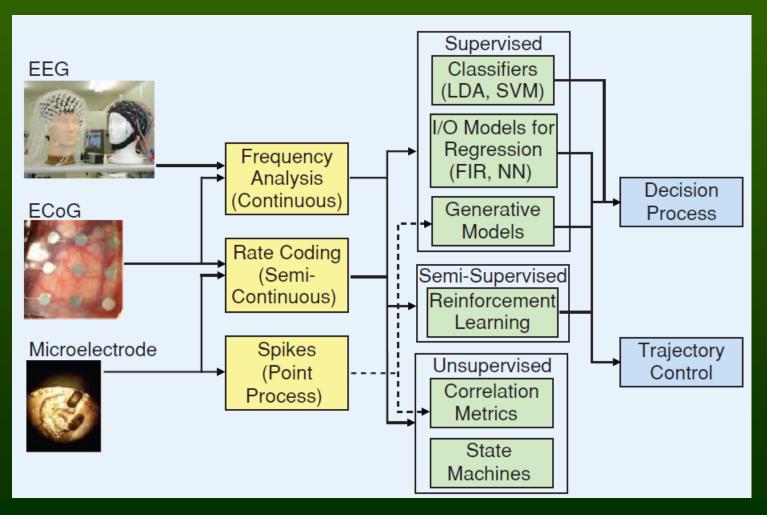




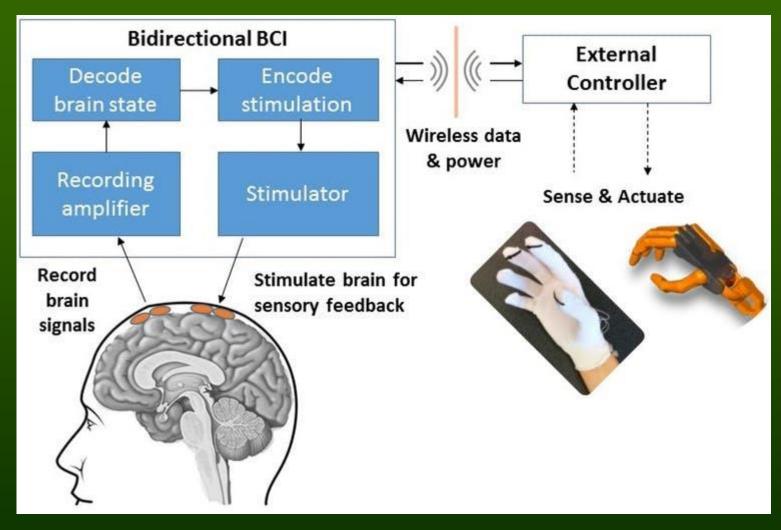
People with Parkinson's disease or compulsive-obsessive disorder who have pacemakers implanted in their brain can regulate their behavior with an external controller.

BCI: time to connect our brains ...

Non-invasive, partially invasive and invasive methods carry increasing amount of information, but are also more difficult to implement. EEG+ML still reigns supreme!



BCBI: Brain-Computer-Brain



BCI + brain stimulation = BCBI – a closed loop through which the brain begins to restructure itself. The body can be replaced by signals in Virtual Reality.

- NONINVASIVE

BCI UNIVERSE

A mind map of sensing and stimulating brain technologies*





MEG

Magneto-Electroencephalography 2 1000c of presented

Major Application	
SQUENTIES	- GP-VSGC
Epilepsy	- Silvey and concentration studies
Strain	- Consumer use for arounal, intention,
Tile	emotion, learning, memory

fMRI



ECoG

Major Applications	
- Epilepsy disgnostics - Speech and movement	- Spinsticotd injury - Locked-in Syndrome - Wovernest disorders
synthesis from neutral decoding	- Movement discordens

- High special coverage



lajor Applications		1 160	è
Otronic pain	- Essential tremor	-145	
Clauser headache	- CICO	-Parkinsons	
Oyosonia	- Huntington's	-Substance Abuse	
Galegoy	- Major depression	-Till	

Implanted Microelectrodes



9	-Looke in Syndrome	·Pe
remonitarila injuly depay	- Movement disorders - 000	3

STIMULATING

BRAIN

Optogenetics

- High spatial and temporal resolution

VNS

Vagus Nerve Stimulation
Noninvasive: tVNS or nVNS

Š	Major Applications:			
	- Abtements - Cancer - Charakteries	- Depresson - Epilepsy - Migranes	- Parumeoris - PTSD	-2004



The second second	Samuel Committee of the
ALS	- Reference and the second
Chronic puin	- Sivep disorders
Computer control Consumer welfness	Stroke rehabilitation Widely used in disprostice



ELECTRO-MAGNETIC

BRAIN

SENSING

fNIRS

of the Sucusion in	sed as a biomarker t	lar besie schild
Major Applications	300	SOUND.
- Destroce - Depression	- Motor execution - Motor imagery	- States - Tild
- Depression - Mental uthreesic	Nusic imagery Nusic imagery	THE

fTCD/tFUS *** Transcranial Focused Ultrasound Stimulation METABOLIC

This is a living document. See updated version at: brainmind.org/bci



tES (tACS, tDCS)

Arretycols		+6/8d T06	
· Alcheimer's		- Parlameants	
Consumer	ORY NAME	- Stroke	
2 27	e imaging	- Skep - Submonor shu	

- · Facy to use

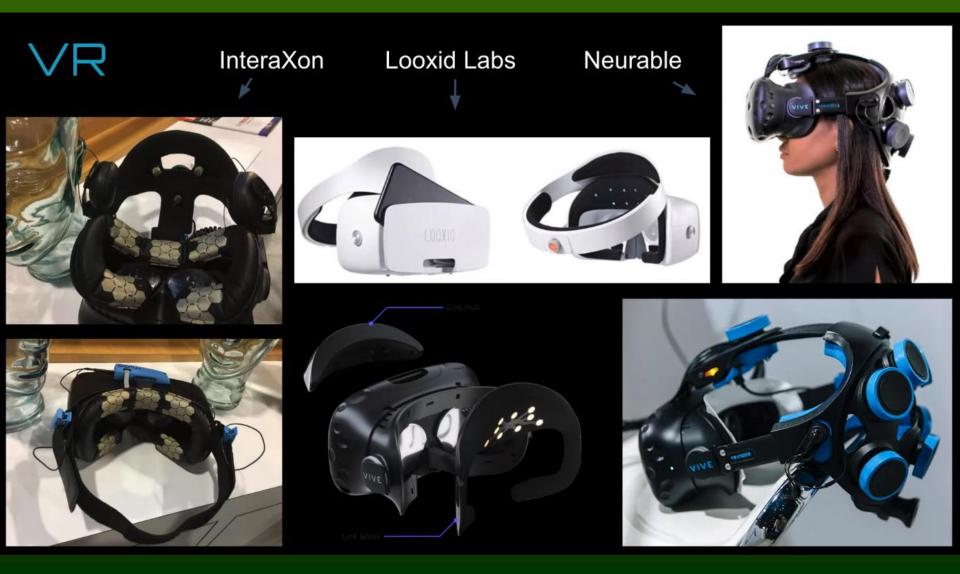
rTMS





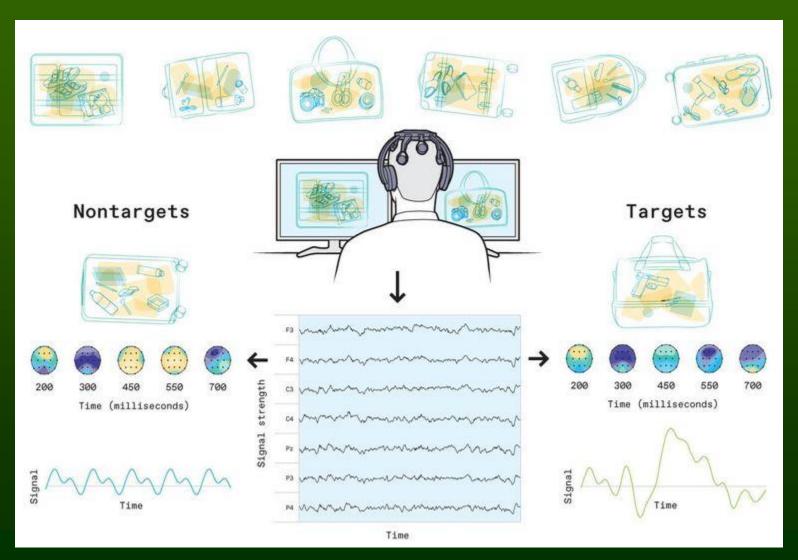
BCI tools

Combination of Virtual Reality with BCI has great potential.



InnerEye

Simple DCS will let you analyze 3-10 images per second for hours ... Your brain is smarter than you!

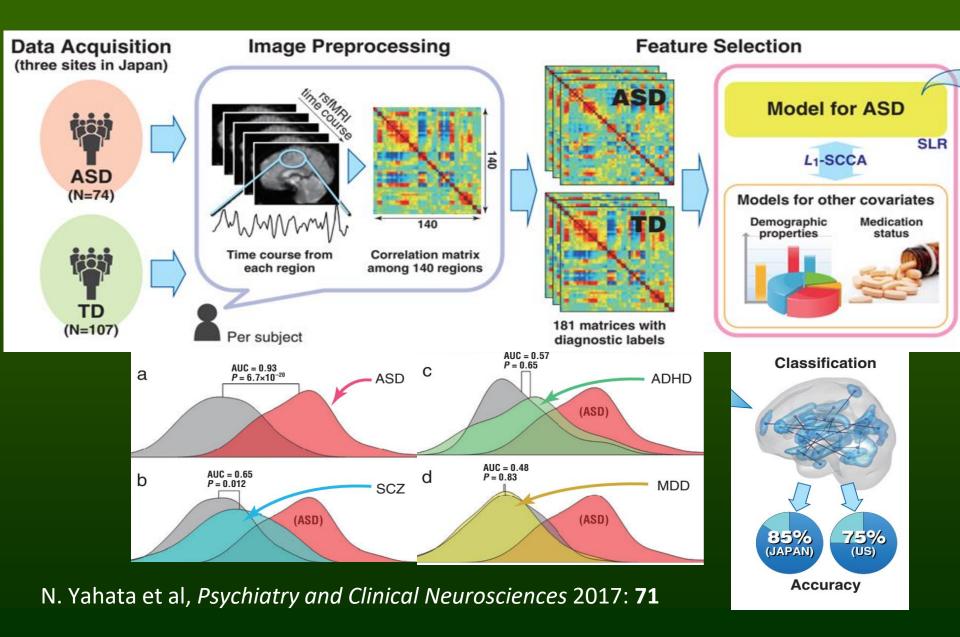


Private Information in BCIs

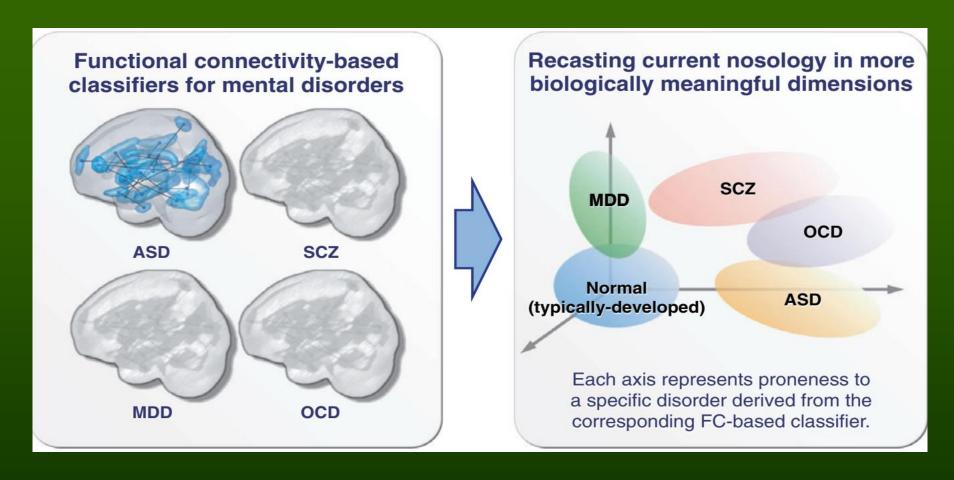


K. Xia, W. Duch, Y. Sun, K. Xu, W. Fang, H. Luo, Y. Zhang, D. Sang, D. Wu, X. Xu, F-Y Wang, <u>Privacy-Preserving Brain-Computer Interfaces</u>: A Systematic Review, IEEE Trans. on Computational Social Systems, 2022

Biomarkers from neuroimaging



Biomarkers of mental disorders



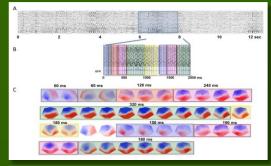
MDD, deep depression, SCZ, schizophrenia, OCD, obsessive-compulsive disorder, ASD autism spectrum disorder. fMRI biomarkers allow for objective diagnosis.

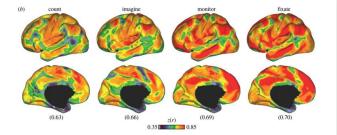
N. Yahata et al, *Psychiatry & Clinical Neurosciences* 2017; **71**: 215–237

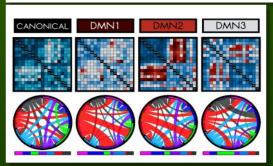
Use it in neurofeedback.

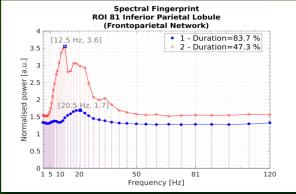
Brain fingerprinting

- Find unique patterns of brain activity, identify:
 - brain regions of interest (ROI)
 - active neural networks
 - mental states, tasks, processes.
- Several approaches:
- 1. Microstates and their transitions (Michel & Koenig 2018)
- 2. Reconfigurable task-dependent modes (Krienen et al. 2014)
- 3. Contextual Connectivity (Ciric et al. 2018)
- 4. Spectral Fingerprints (Keitel & Gross 2016)
- 5. fMRI networks (Yuan ... Bodurka, 2015).
- 6. Recurrence quantification analysis.
- + many more approaches...











Understanding brains: microstates

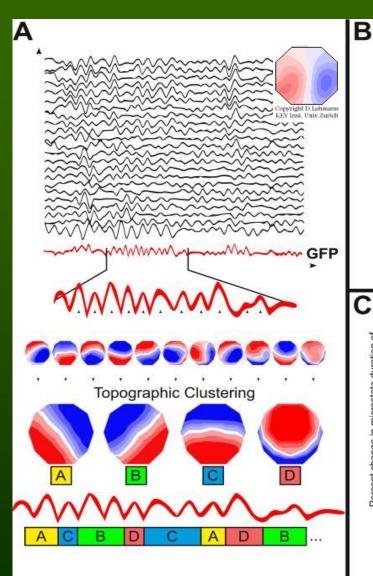
Global Field Power. 4-7 states, 60-150 ms.

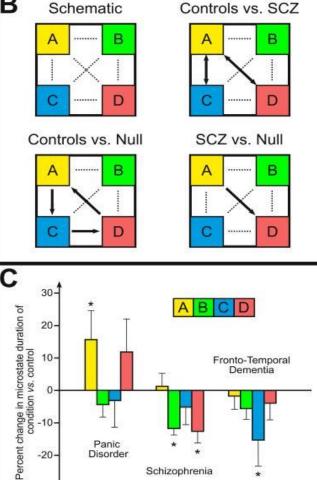
Khanna et al. (2015) Microstates in Resting-State EEG. Neuroscience and Biobehavioral Reviews.

Symbolic dynamics:

statistics of A-D symbol strings. Fuzzy Symbolic Dynamics (FSD) + visualizations.

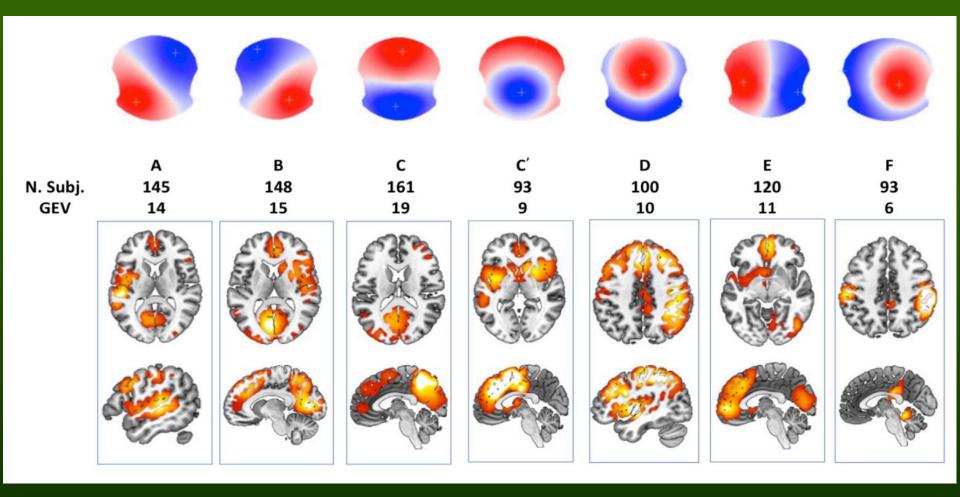
Duch W, Dobosz K. (2011). Cognitive Neurodynamics 5, 145 Dobosz K, Duch W. (2010). Neural Networks, 23(4), 487-496.





Schizophrenia

Microstates and their sources



Michel, C. M., & Koenig, T. (2018). EEG microstates as a tool for studying the temporal dynamics of whole-brain neuronal networks: A review. *NeuroImage*, 180, 577–593. Ewa Ratajczak, PhD thesis "Microstate neurodynamics in HRV biofeedback" (2022)

Spectral fingerprints of cognitive processes

Decompose neurodynamics.
Find subnetworks binding ROIs at specific frequencies.
Oscillations can rapidly change, one ROI is engaged in different subnetworks for short time periods.
This is reflected very crudely in microstates, recurrence plots show more precise information.

Siegel, M., Donner, T. H., & Engel, A. K. (2012). Spectral fingerprints of largescale neuronal interactions. *Nature Reviews Neuroscience*, *13*(2), 121–134.

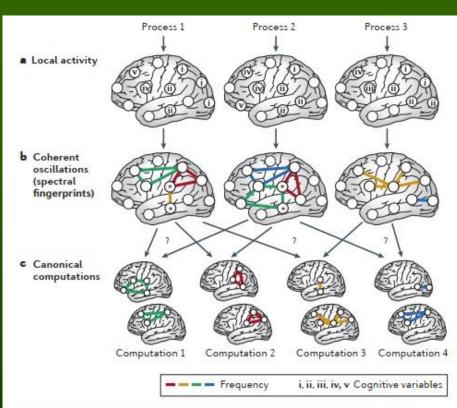
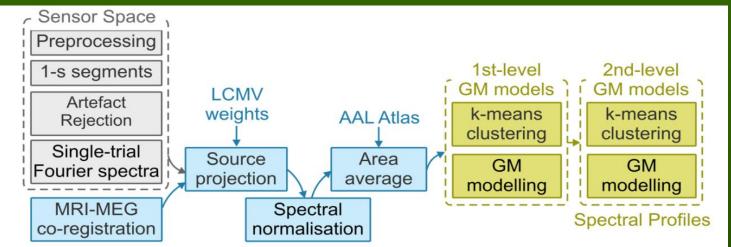


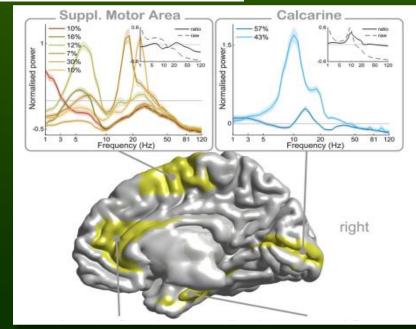
Figure 4 | Large-scale spectral fingerprints of cognitive processes. Schematic illustration of how coherent oscillations provide 'spectral fingerprints' for regrouping of cognitive processes 1–3. a | Studies of neuronal activity in individual brain regions (circles) elucidate the activation of different regions (bold circles) and the encoding of various cognitive variables (Roman numerals) during different cognitive processes. Several cognitive variables (for example, different sensory features) are simultaneously encoded in each region, but for simplicity only one variable is depicted per region. Note that the pattern of local activity and encoding can be similar between processes. b | Coherent oscillations allow for the characterization of the interactions between different brain regions (coloured lines) during different cognitive processes. The frequency of these oscillations (indicated by the colours) allows the corresponding network

Spectral analysis

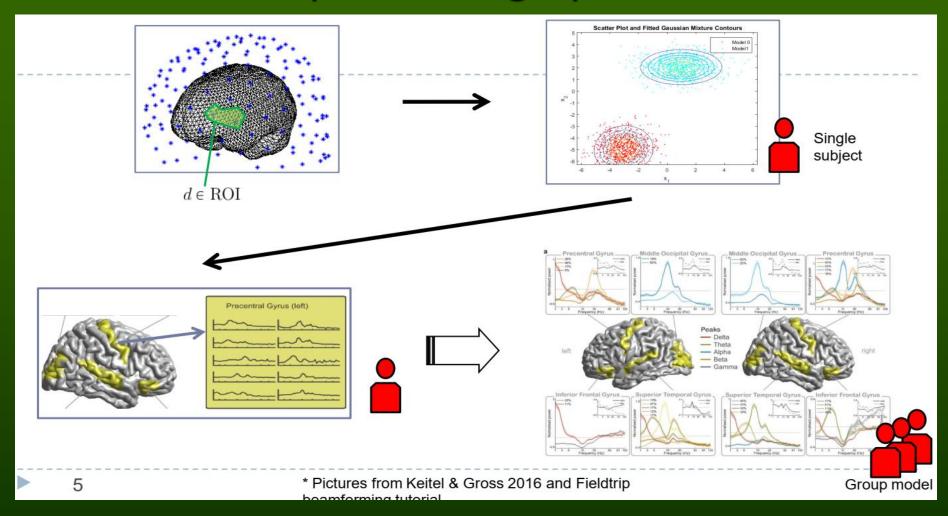


Create spectral fingerprints of ROIs.

- Analyze EEG/MEG power spectra in 1 sec time windows; project them to the source space of ROIs based on brain atlas; clusterize individual/group to create spectra.
- A. Keitel & J. Gross. Individual human brain areas can be identified from their characteristic spectral activation fingerprints. PLoS Biol 14, e1002498, 2016



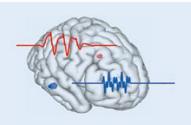
Spectral fingerprints



One ROI, two or more spectra. Static picture showing natural frequencies. Michał Komorowski PhD "Locally specific human brain dynamics automatically modeled using spectral features of MEG/EEG signals", 2022. ToFFi toolbox.

EEG localization and reconstruction

ECD



$$\widehat{d_j} = \operatorname{argmin} \parallel \phi - \sum_j \mathcal{K}_j d_j \parallel_{\mathcal{F}}^2$$

Rotating dipole

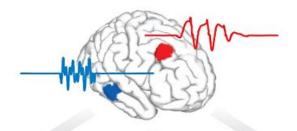
- Moving
- Rotating
- Fixed

Dipole model



Distributed model



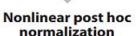


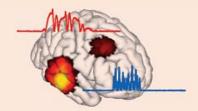
MN (ℓ_2) family



$$\mathbf{\hat{\jmath}} = \underset{\mathbf{j}}{\operatorname{argmin}} \| \boldsymbol{\phi} - \mathcal{K} \mathbf{j} \|_{2}^{2} + \lambda \| \mathbf{j} \|_{2}^{2}$$
$$\mathbf{\hat{\jmath}} = \mathcal{T} \boldsymbol{\phi} = \mathcal{K}^{\mathsf{T}} (\mathcal{K} \mathcal{K}^{\mathsf{T}} + \lambda \mathbf{I})^{\mathsf{T}} \boldsymbol{\phi}$$

- LORETA · MN
- WMN



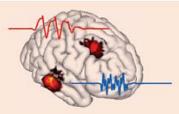


$$\hat{\boldsymbol{j}}_{mn} = \boldsymbol{T}_{mn} \boldsymbol{\phi}$$

$$\boldsymbol{S}_{\boldsymbol{j}} = \boldsymbol{\mathcal{K}}^{\mathsf{T}} (\boldsymbol{\mathcal{K}} \boldsymbol{\mathcal{K}}^{\mathsf{T}} + \alpha \boldsymbol{I})^{\dagger} \boldsymbol{\mathcal{K}}$$

$$\hat{\boldsymbol{j}}_{SL} = \hat{\boldsymbol{j}}_{mn} (\boldsymbol{\ell})^{\mathsf{T}} \left([\boldsymbol{S} \hat{\boldsymbol{j}}]_{\boldsymbol{\ell} \boldsymbol{\ell}} \right)^{-1} \hat{\boldsymbol{j}}_{mn} (\boldsymbol{\ell})$$
SLORETA

He et al. Rev. Biomed Eng (2018) Beamforming and

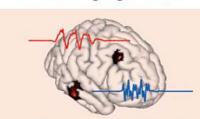


Sparse and

Bayesian framework

$$\begin{aligned} \hat{\mathbf{\jmath}} &= \underset{j}{\operatorname{argmin}} \parallel \mathcal{V}_{\dot{\mathbf{\jmath}}} \parallel_{1} + \alpha \parallel \mathbf{\jmath} \parallel_{1} \\ &\text{S.T.} \parallel \phi - \mathcal{K}_{\dot{\mathbf{\jmath}}} \parallel_{\Sigma^{-1}}^{2} \leq \varepsilon^{2} \end{aligned}$$

IRES



scanning algorithms

$$\widehat{\boldsymbol{w}}_{r} = \underset{\boldsymbol{w}_{r}}{\operatorname{argmin}} \ \boldsymbol{w}_{r}^{\mathsf{T}} \mathcal{R}_{\phi} \boldsymbol{w}_{r}^{\mathsf{T}}$$
S.T.
$$\begin{cases} \mathcal{K}_{r}^{\mathsf{T}} \boldsymbol{w}_{r} = \boldsymbol{\xi}_{1} \\ \boldsymbol{w}_{r}^{\mathsf{T}} \boldsymbol{w}_{r} = \boldsymbol{1} \end{cases}; \boldsymbol{j} = \boldsymbol{w}^{\mathsf{T}} \phi$$

Beamformer (VBB)

SupFunSim

- SupFunSim: our library/Matlab /tollbox, direct models for EEG/MEG, on GitHub.
- Provides many spatial filters for reconstruction of EEG sources: linearly constrained minimum-variance (LCMV), eigenspace LCMV, nulling (NL), minimum-variance pseudo-unbiased reduced-rank (MV-PURE) ...
- Source-level directed connectivity analysis: partial directed coherence (PDC), directed transfer function (DTF) measures.
- Works with FieldTrip EEG/ MEG software. Modular, object-oriented, using Jupyter notes, allowing for comments and equations in LaTex.

K. Rykaczewski, J. Nikadon, W. Duch, T. Piotrowski, *Neuroinformatics* **19**, 107-125, 2021.

end

STFT vs. embedding

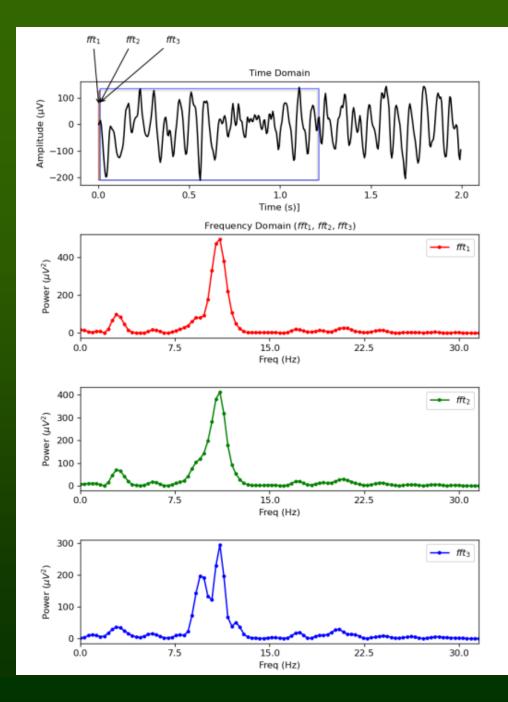
Takens theorem: attractors are recreated from signals sampled using time-delay embedding, vectors

 $\boldsymbol{x}_i = (u_i, u_{i+\tau}, ..., u_{i+(m-1)\tau\Delta t}).$

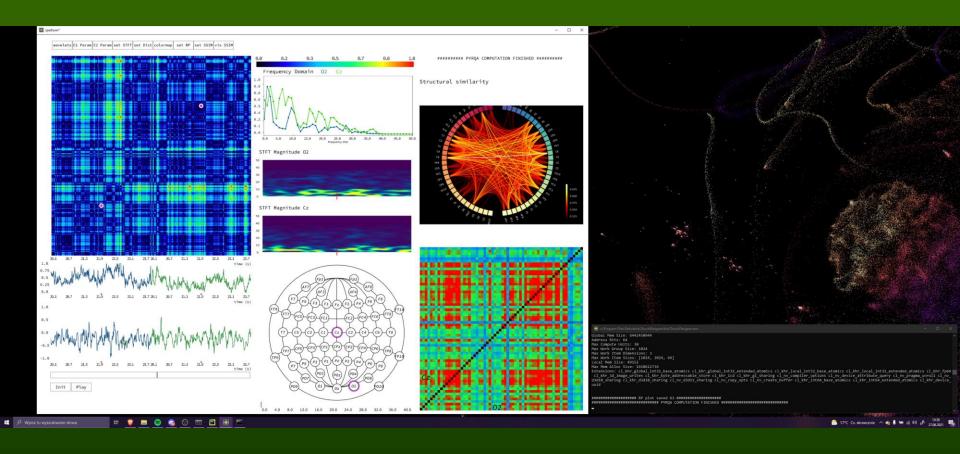
Here m is the embedding dimension, and τ is an index enumerating time delays, $\tau \Delta t$.

Alternative representation: STFT, shows power distribution in subsequent time windows. Here changes of spectrum every 100 ms, O_1 electrode.

Ł. Furman, W. Duch, L. Minati, K. Tołpa, Short-Time Fourier Transform and Embedding Method for Recurrence Quantification Analysis of EEG Time Series. The European Physical Journal Special Topics (2022, p. 1-15).

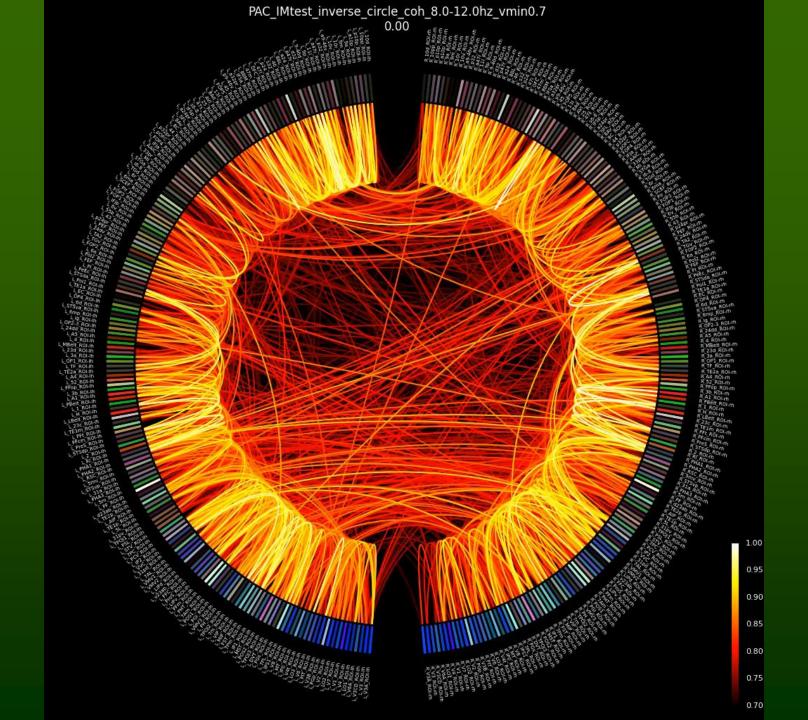


STFT EEG in real time

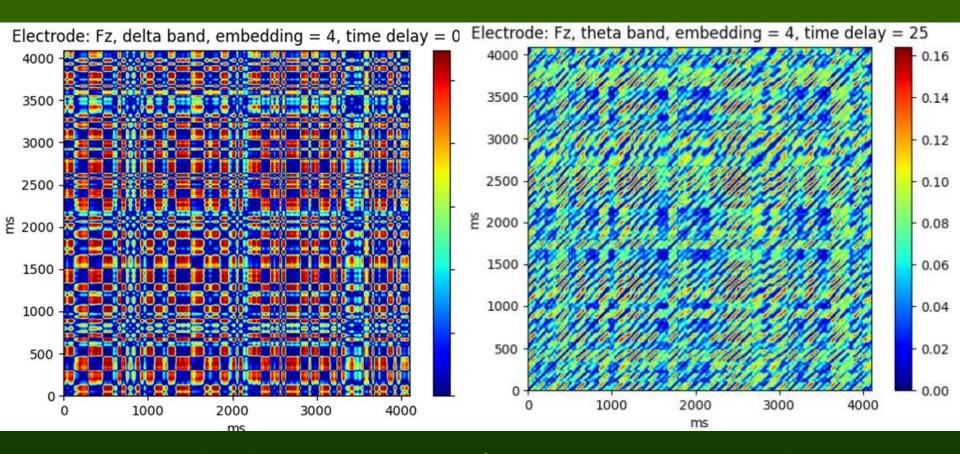


EEG data, 128 channels, recursion graphs, power spectrum for two electrodes, information flow and correlations between brain regions.

BrainPulse (in development, Łukasz Furman).



Recurrence plots δ , θ



Recurrence Plots showing STFT similarity for delta and theta bands, Fz electrode. Distance scale changes parameters of the metastable states along diagonal, and influence non-linear parameters. BrainPulse tools helps to analyze changes of the t/f spectra, RPs and STFT power spectra.

Labeling states

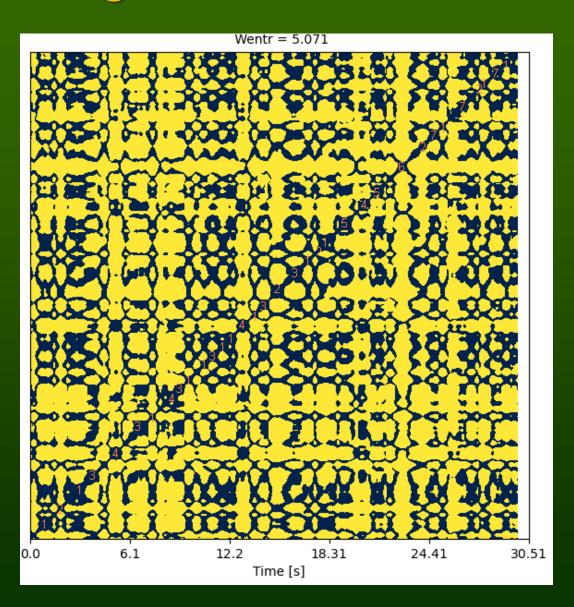
Automatic labeling of states and estimation of their recurrence may be important for biofeedback.

Metabolic costs of transitions between states may be important.

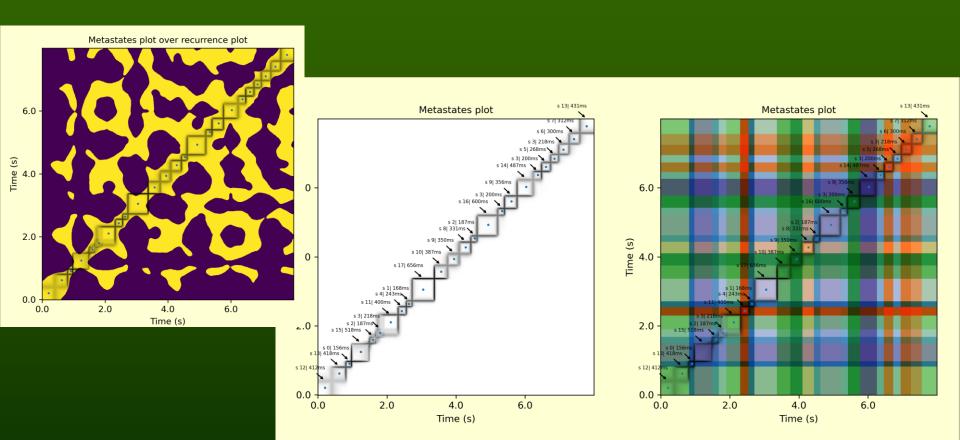
Ruminations? Pain states? How external stimuli influence this dynamics?

Needs automatic method for recognition of metastable, multivariate states.

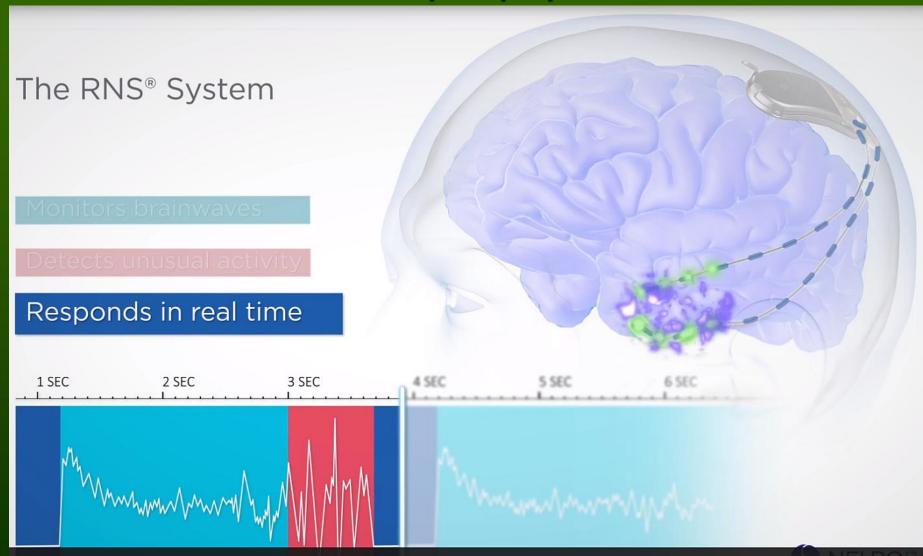
More precise than microstates.



Segmentation of states



Epilepsy



The neurostimulator and detector stops attacks of drug-resistant epilepsy before cramps occur. About 1% of people in the world have epilepsy.

Brain Robotic Interface

<u>Australia, UTS</u>: VR to control robotic dogs using EEG.
 Dry graphene sensors, not as accurate as wet. Can it be useful?





A million nanowires in the brain?

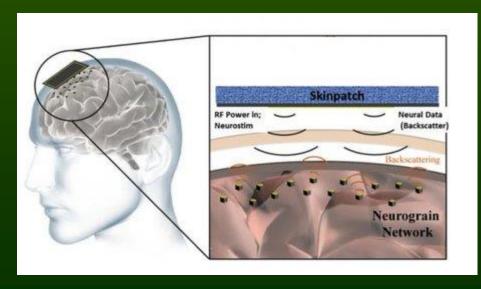
DARPA initiatives: **Neural Engineering System Design (NESD)** and other projects.

An interface that reads the impulses of 10⁶ neurons, stimulates 10⁵ neurons, simultaneously reads and stimulates 10³ neurons.

DARPA awarded grants to research groups for projects under the program <u>Electrical Prescriptions (ElectRx)</u>, whose aim is to develop BCBI systems modulating the activity of peripheral nerves for therapeutic purposes.

Neural dust – microscopic wireless sensors in the brain.

Elon Musk promised neuralink (neural lace). FDA refuses human trials.





Perspectives

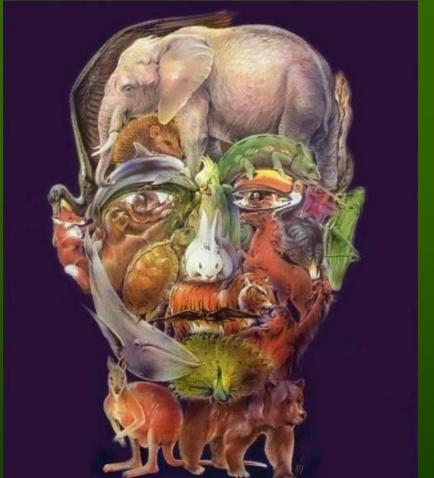
- Al is changing everything, including the way science is done.
- Large language models can do much more than generate language and images.
- AutoGTP can plan and recruit many tools, performing complex tasks.
- Such techniques have a chance to improve brain signal processing, but we still need better methods for understanding brain activity.
- Neurocognitive technologies will profoundly change ourselves, ultimately restoring normal functions and optimizing normal brain processes, but there is a lot of hype in the BCI field.
- Optimization of brain processes is the biggest challenge!

Caution! Singularity may come faster than we think!



Intelligence?





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