

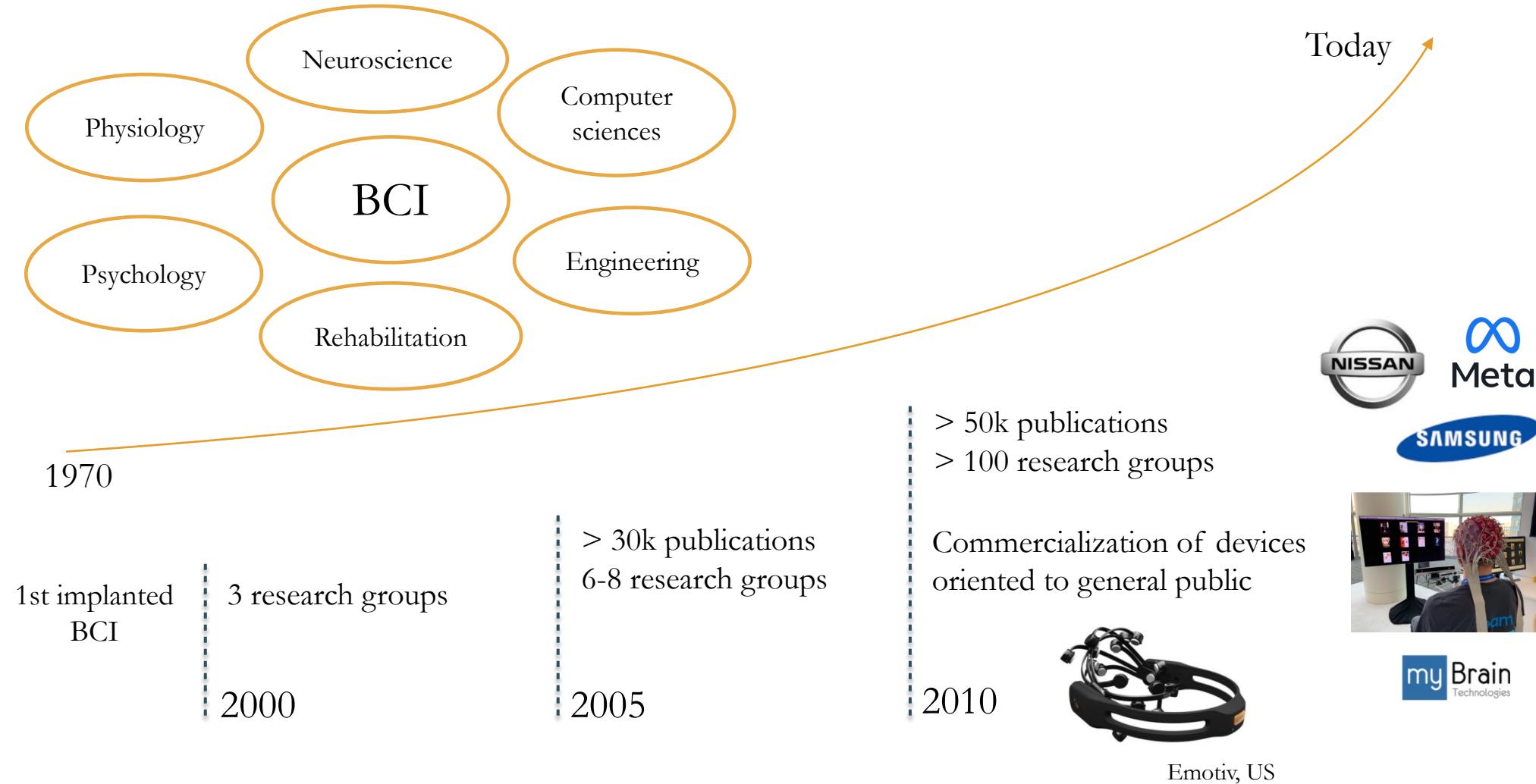
WHAT ARE BCIS, AND HOW DO THEY WORK? A GUIDED TOUR!



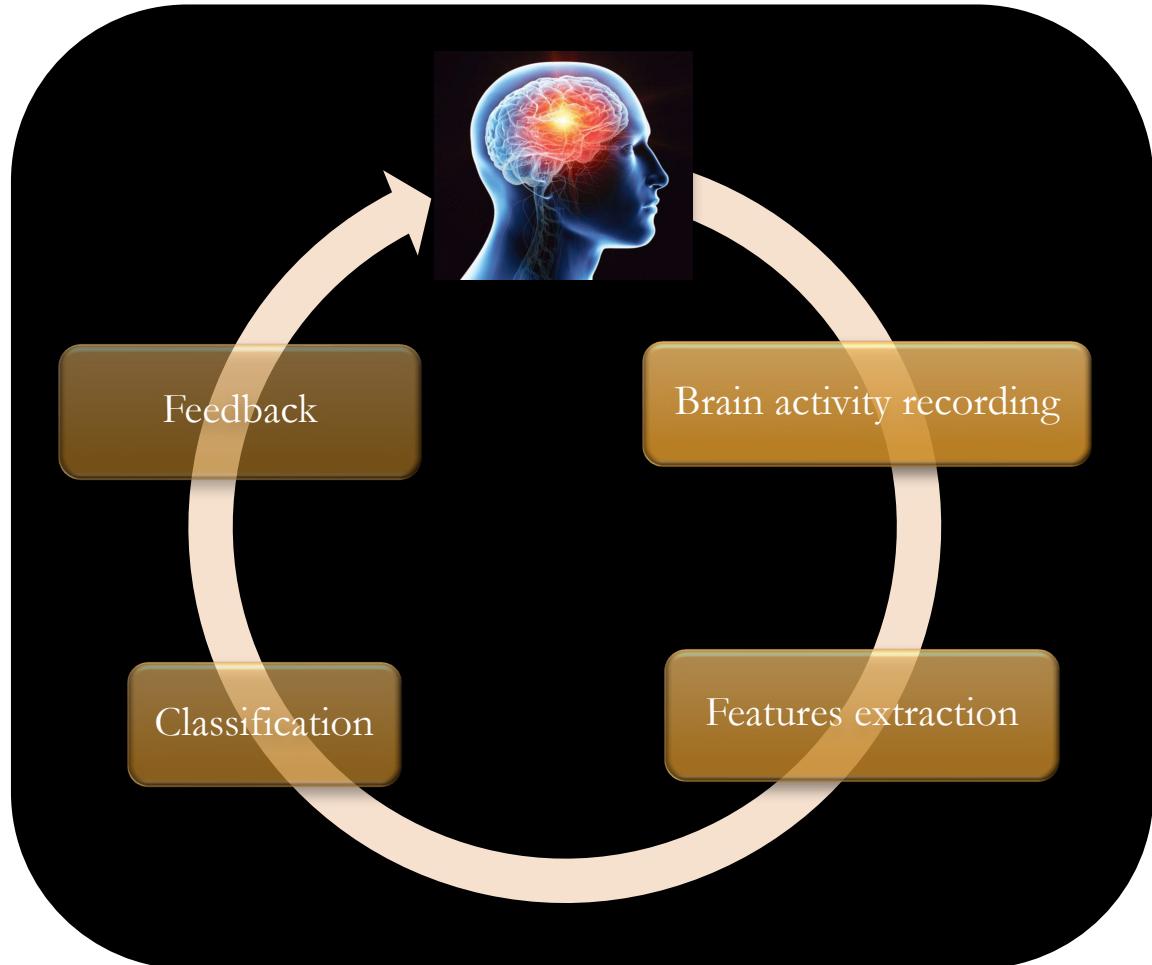
Arthur Desbois & Marie-Constance Corsi
Inria NERV team, Paris Brain Institute



Context

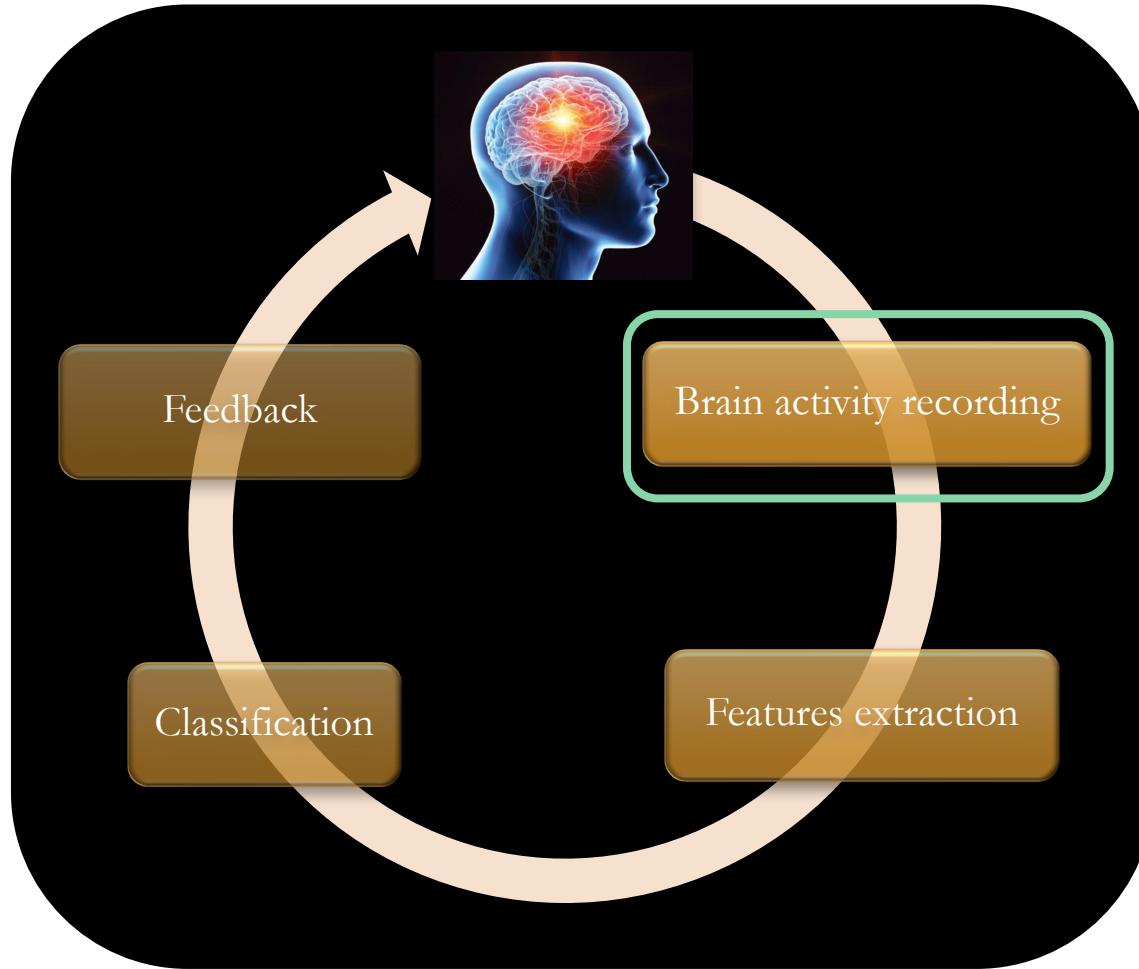


Behind the magic...



Behind the magic...

5



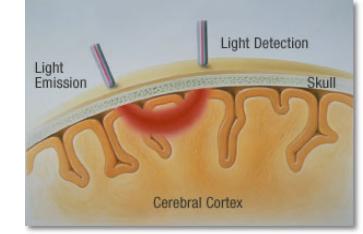
Non-invasive tools



EEG



MEG



NIRS



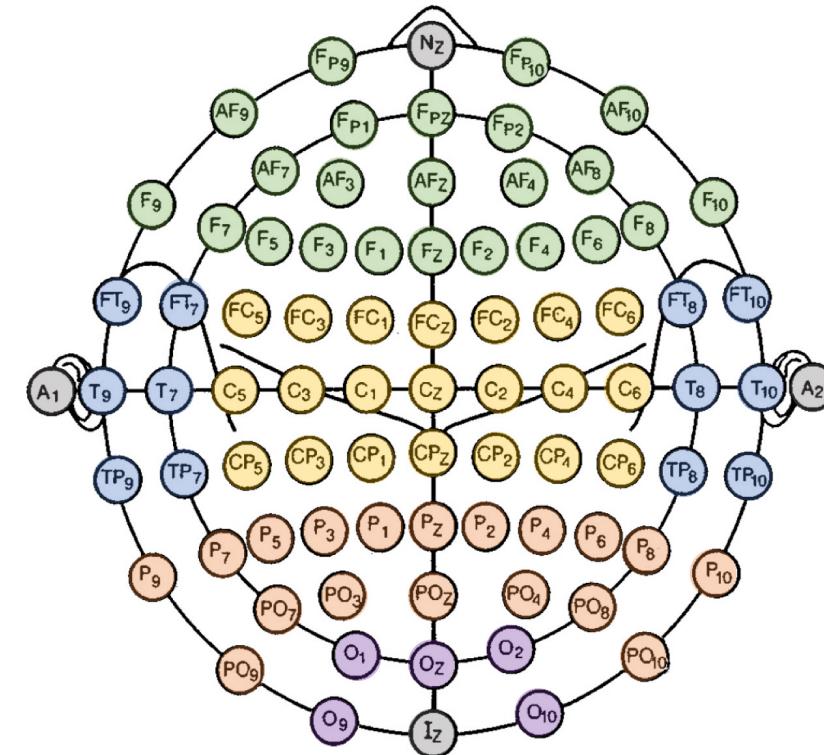
fMRI

Near Infrared Spectroscopy
(Fazli et al, 2012)

Functional MRI
(Sitaram et al, 2009)

Brain recording – EEG instrumentation

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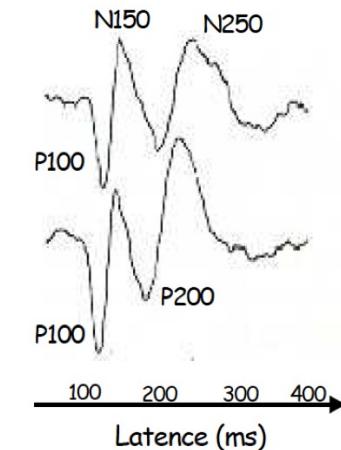
Adapted from (Corsi, 2023)

Brain recording w/ M/EEG – Evoked responses

7

Nomenclature: the latency, the amplitude, the shape and the polarity

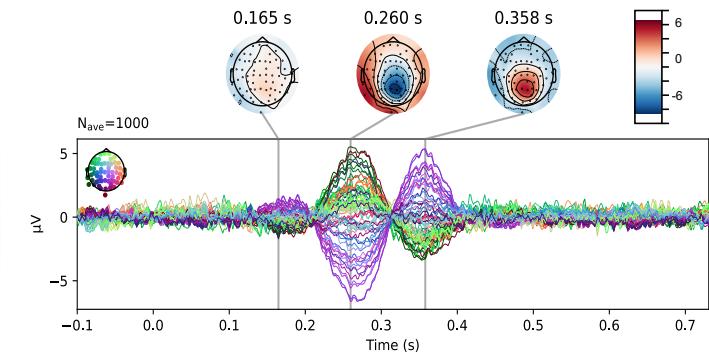
- Nxxx: one negative wave @ xxx ms (EEG)
- Pxxx: one positive wave @ xxx ms (EEG)
- Mxxx: one wave @ xxx ms (MEG)



Adapted from [Campagne, 2014]

Components

- Early components (exogenous): related to stimulus characteristics
- Late components (endogenous): related to the task, to the subject's stat



Adapted from [Corsi, 2023]

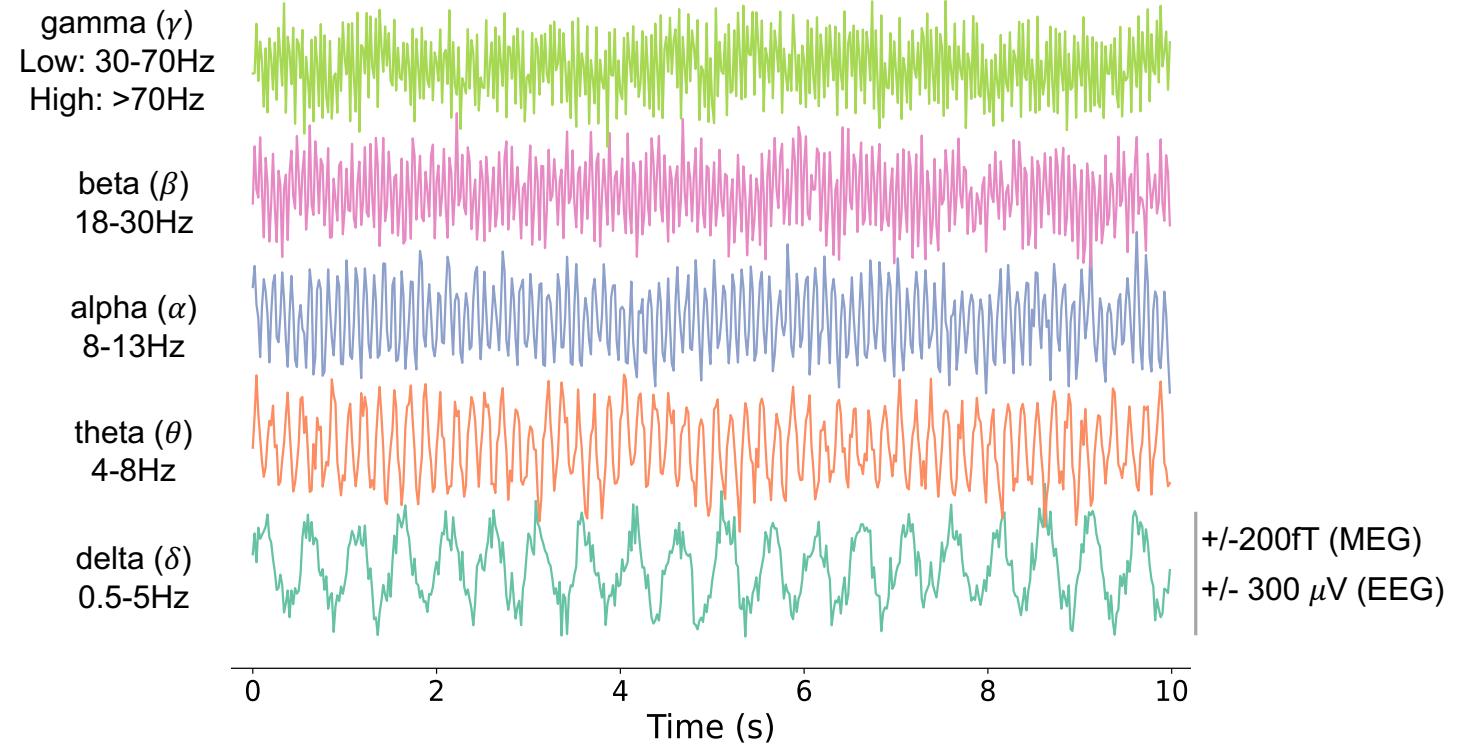
BCI application: P300 Speller

Brain recording w/ M/EEG – Oscillatory activity

8

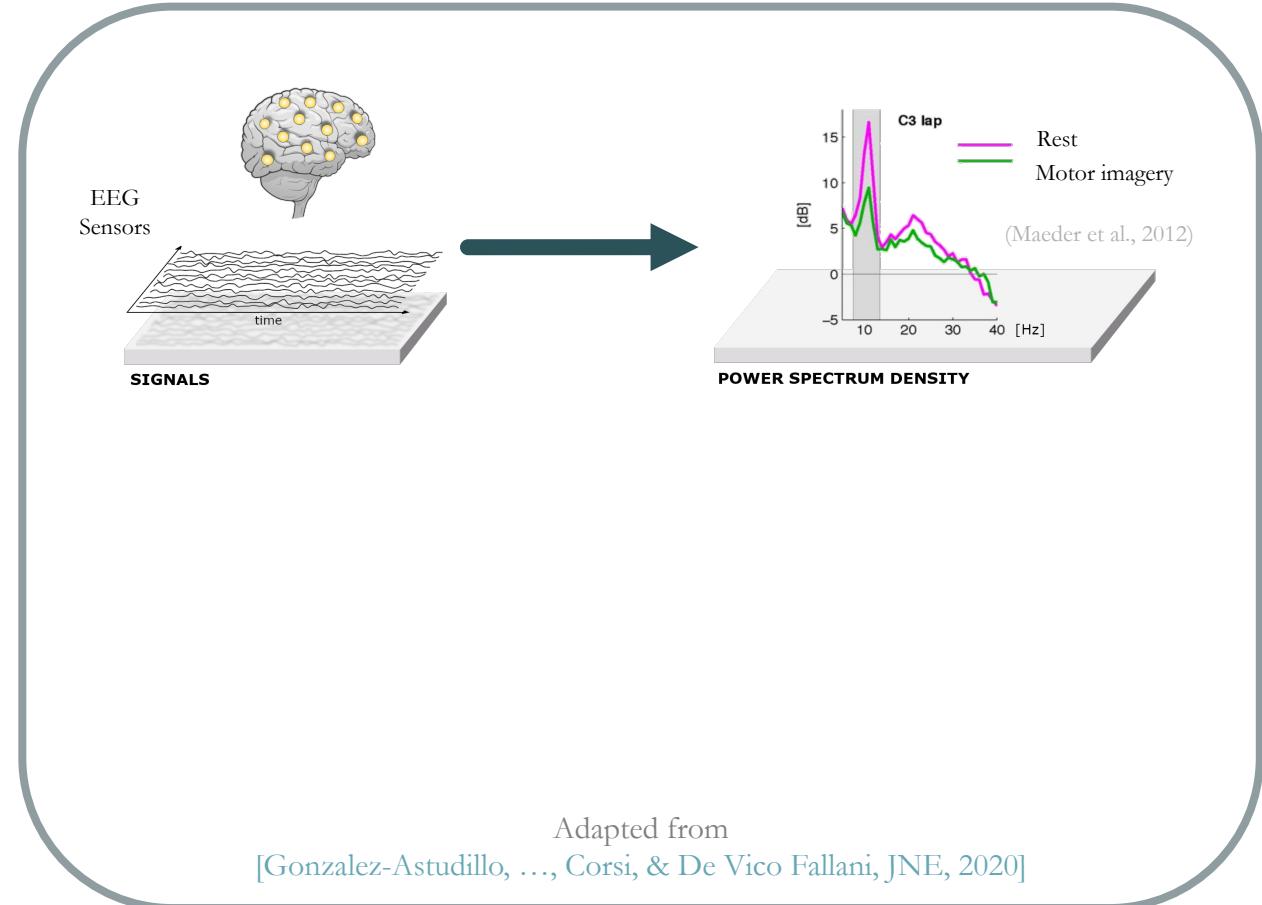
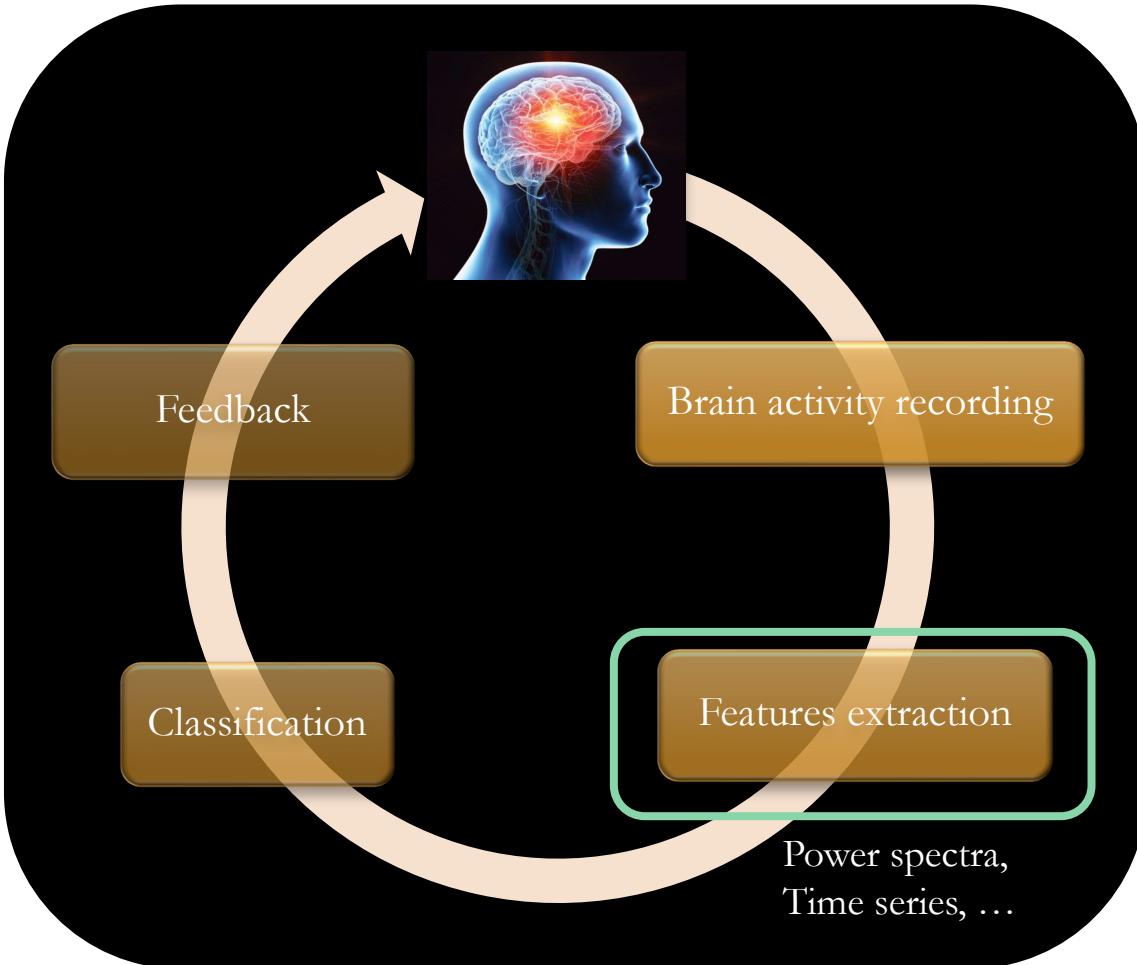
Spontaneous activity, characteristics:

- Frequency
- Amplitude
- Shape
- Localization
- Psychopsychological context
- Duration
- Vanishing



Adapted from [Corsi, 2023]

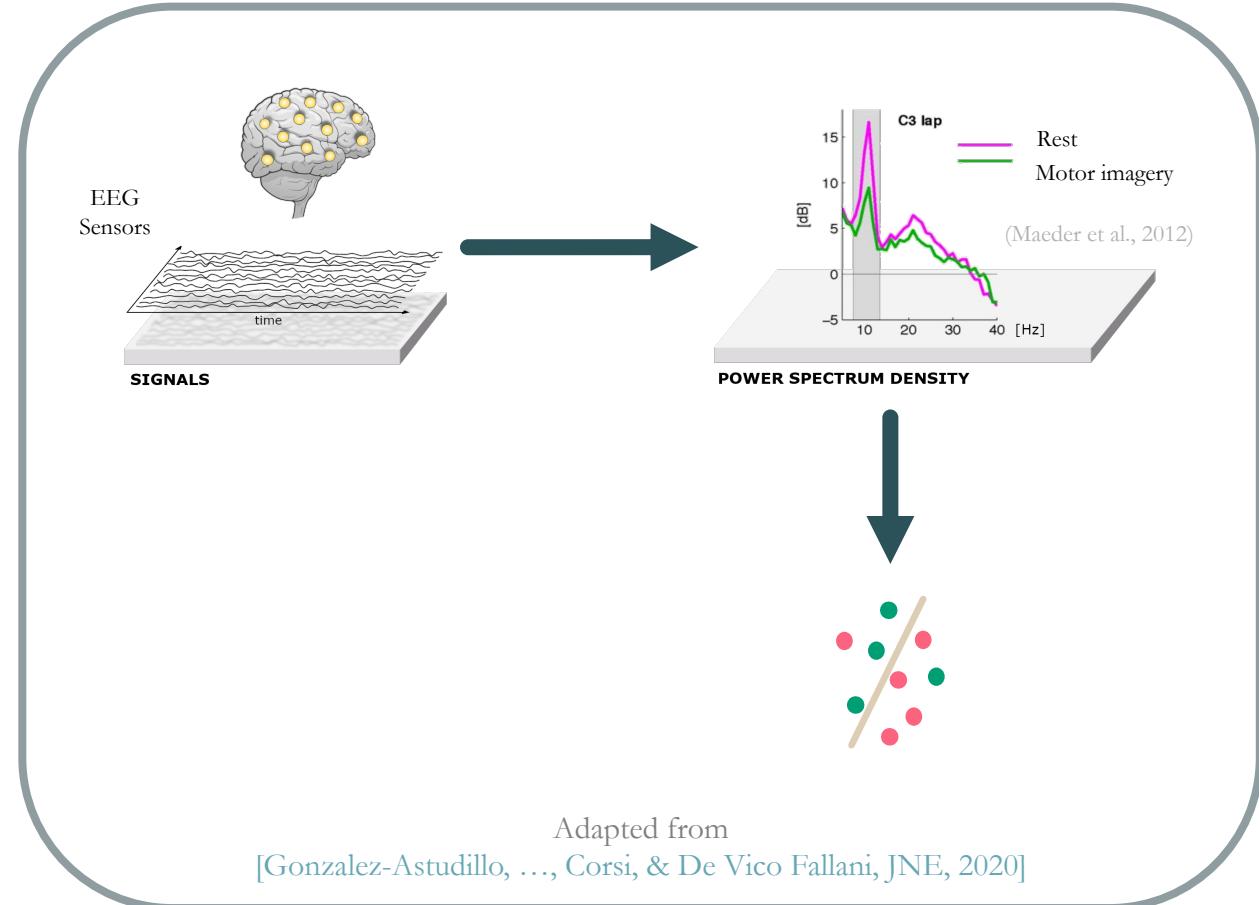
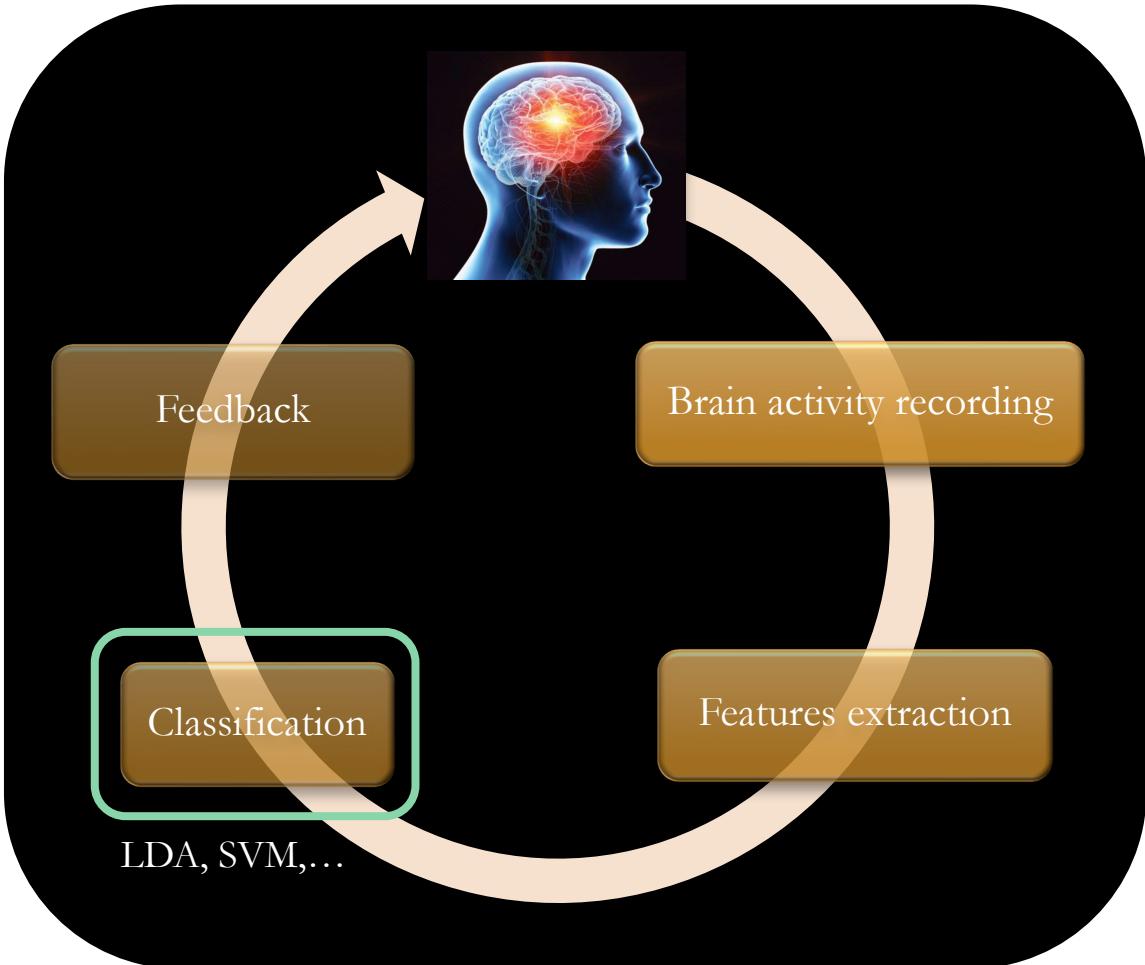
Behind the magic...



Adapted from
 [Gonzalez-Astudillo, ..., Corsi, & De Vico Fallani, JNE, 2020]

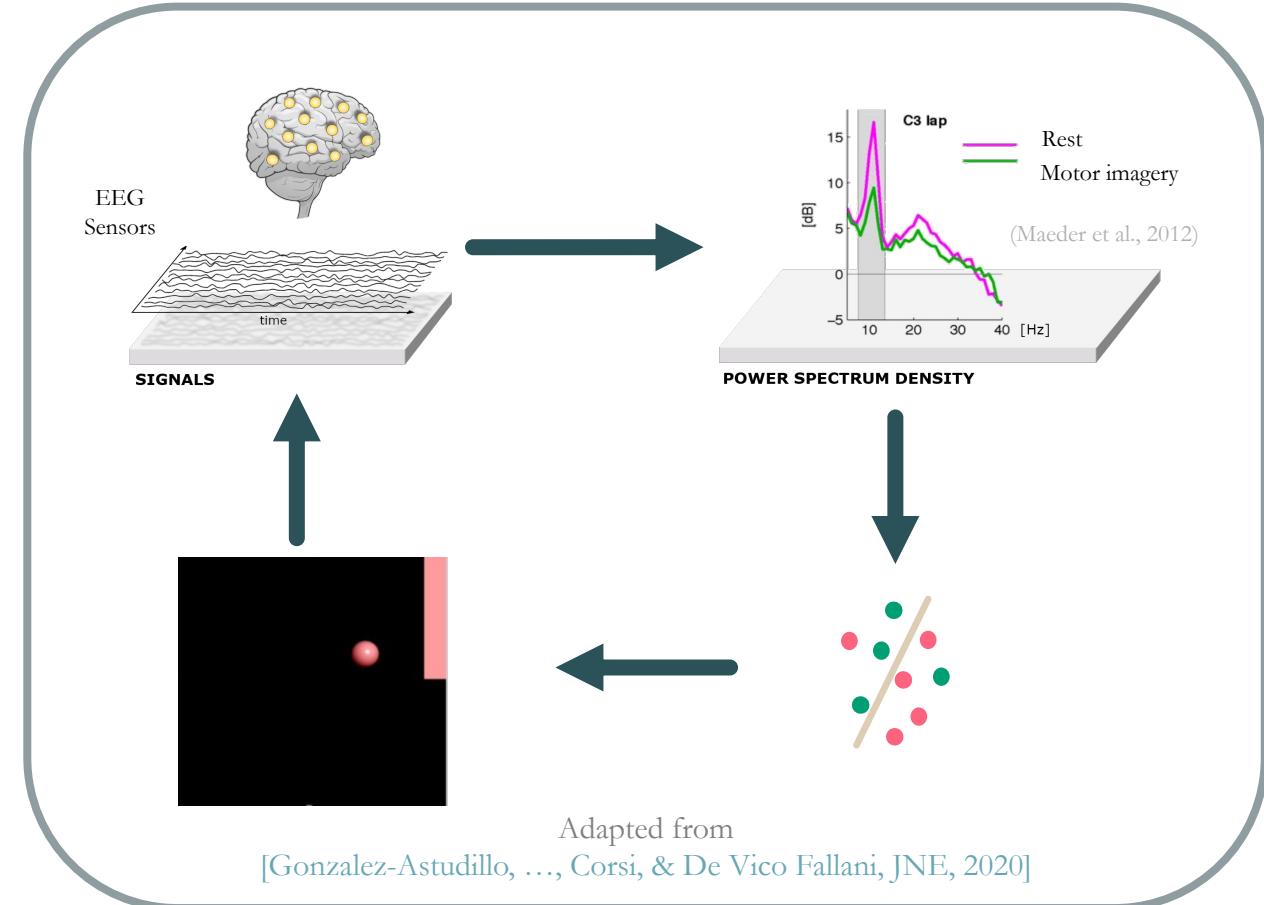
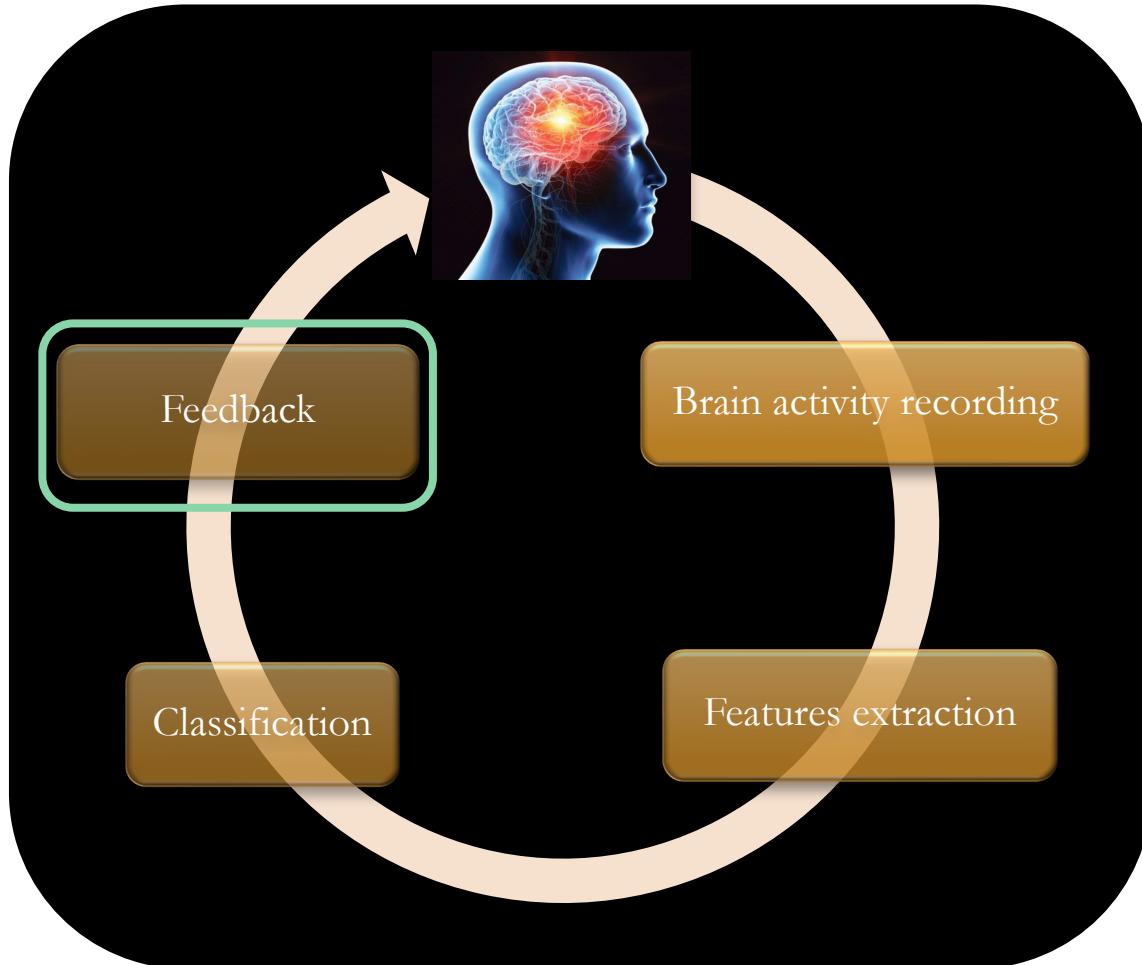
Behind the magic...

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Behind the magic...

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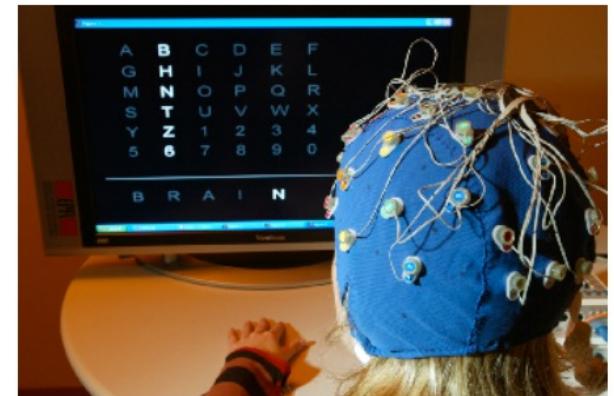
■ Control

- Prosthesis (Fifer et al, 2014)
- Wheelchair (Carlson & Millan, 2013)
- Quadcopter (LaFleur et al, 2013)



■ Communication

- Verbal & nonverbal communication (Jin et al, 2012; Hwang et al, 2012; Kashihara, 2014)
- Silent talk (Naci et al, 2013)

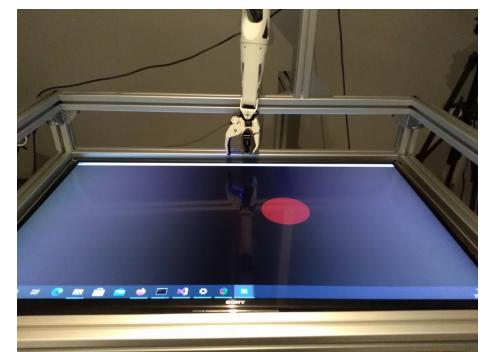
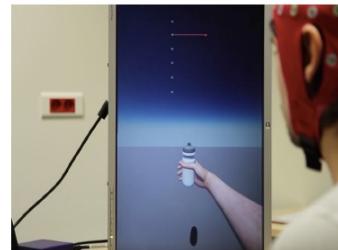


BCI & communication

■ Neurological disorders treatment

- Stroke (Prasad et al, 2010)
- Spinal cord injury (King et al, 2013)
- Consciousness (Chatelle et al, 2012)
- Psychiatric disorders (Arns et al, 2017)

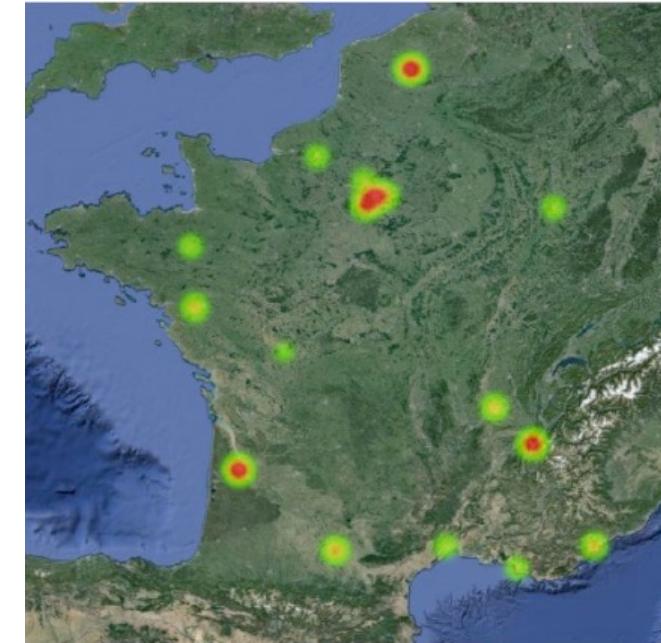
- Inria software development w/ OpenViBE for:
 - Robotic device control
 - Stroke rehabilitation
 - Better monitoring general anesthesia



LORIA projects
Courtesy of S. Rimbert

ARAMIS projects
Courtesy of T. Venot

- Examples of French BCI laboratories
 - LORIA team (Nancy, France)
 - Hybrid team (Rennes, France)
 - Potioc team (Bordeaux, France)
 - NERV team (Paris, France)
- Most salient disciplines:
 - EEG Signal Processing & Machine Learning
 - Clinical Neuroscience
 - Human-Computer Interaction & BCI
 - Computational Neuroscience
 - Invasive BCI research
 - Ethics

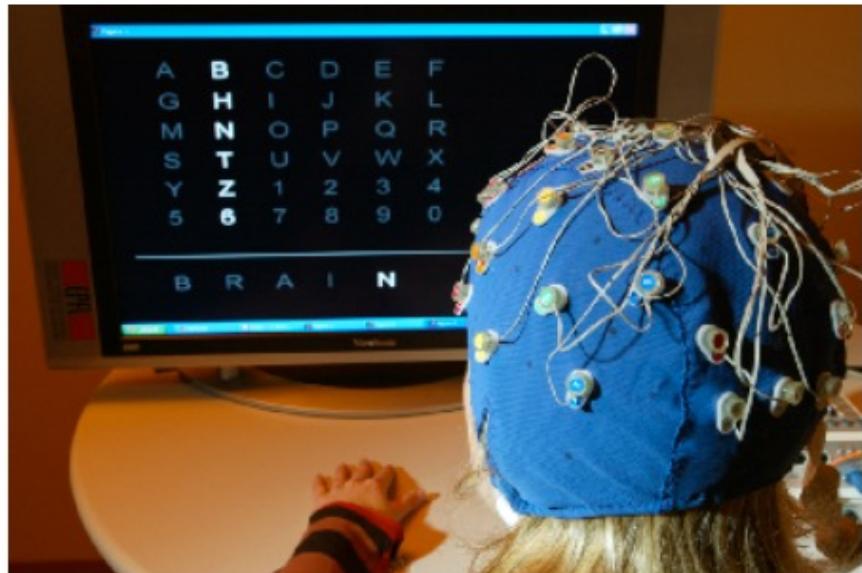


BCI labs localization in France

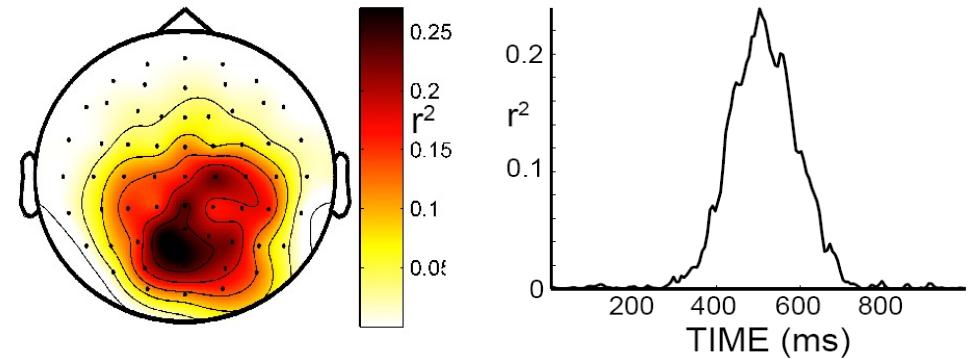
⇒ Access to an interactive map of laboratories: [here](#) (work in progress, not exhaustive!)

Different types of BCI – P300 Speller

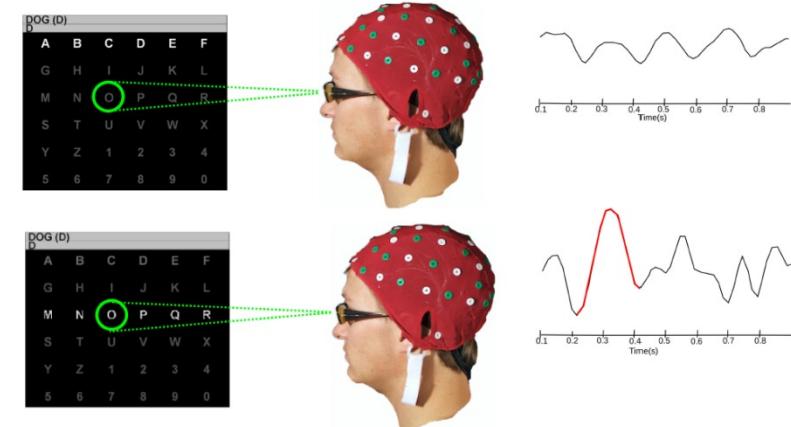
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P300 Speller



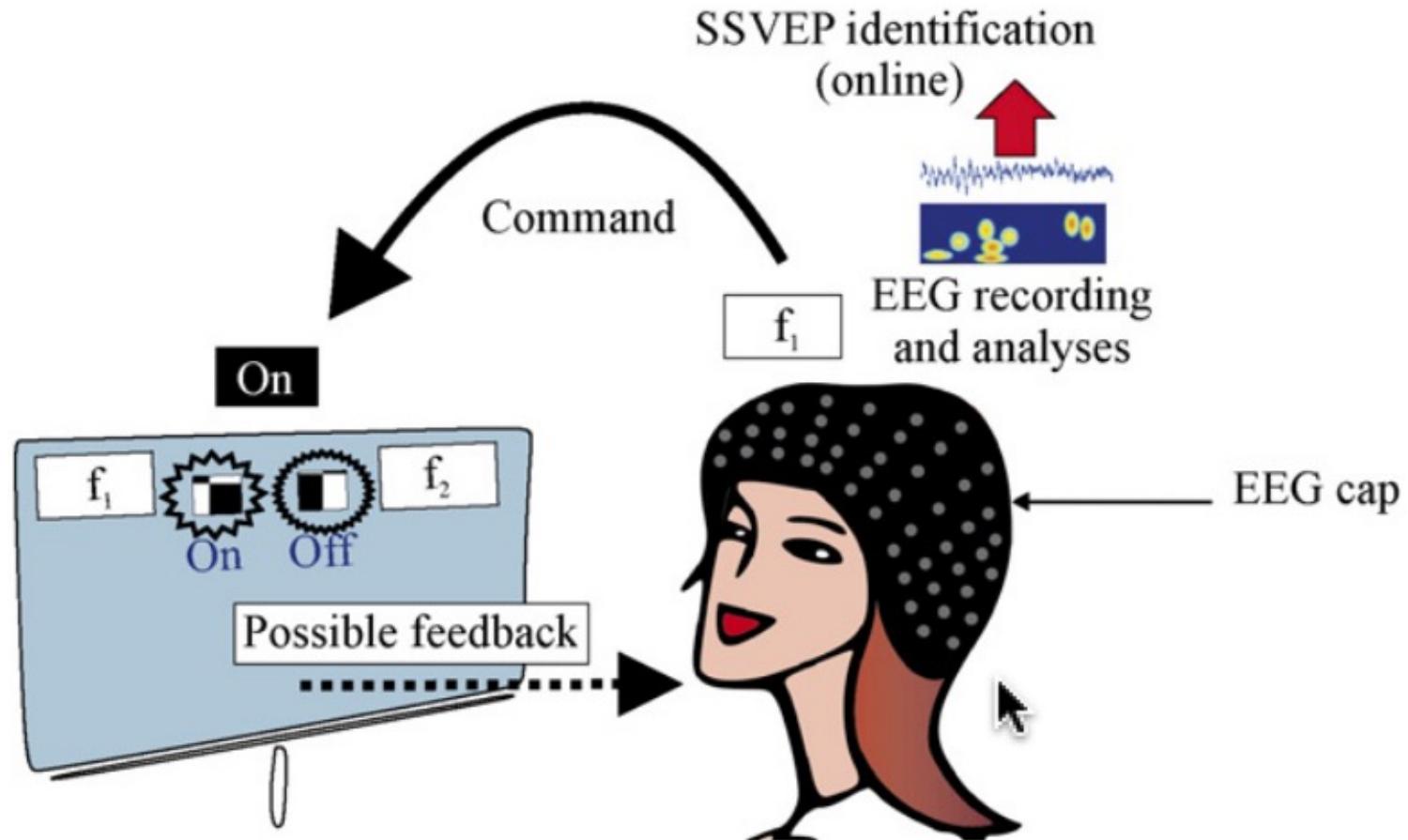
Illustrations from BCI2000 website



Adapted from [Lotte et al, 2015]

Different types of BCI – Visual epoked potential

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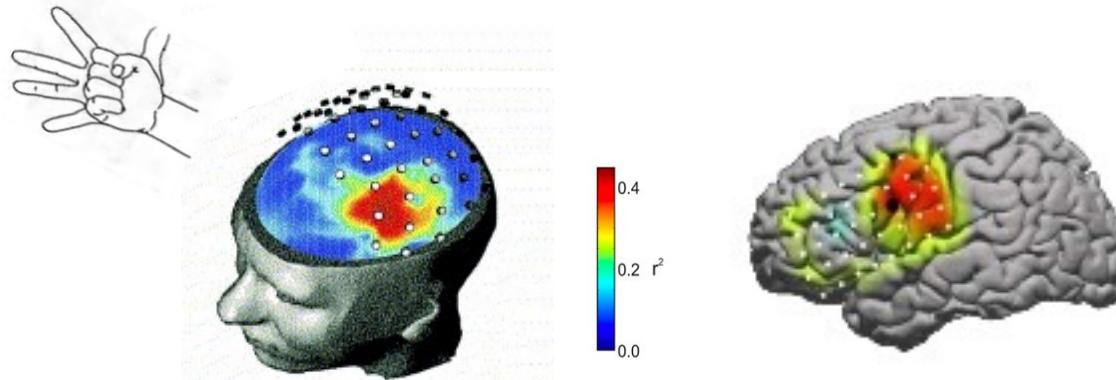


(Vialatte et al, 2010)

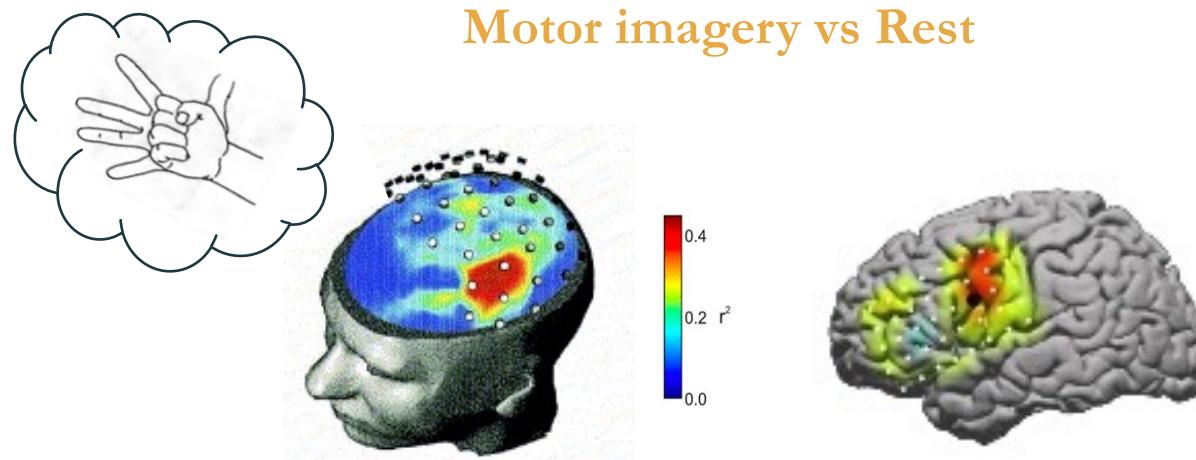
Different types of BCI – Motor imagery

17

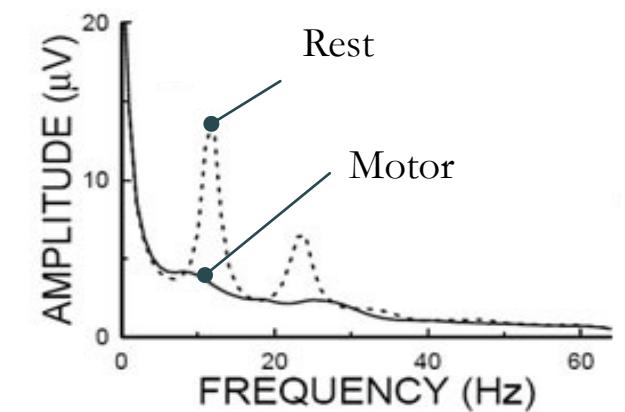
Motor execution vs Rest



Motor imagery vs Rest



Power decrease



Desynchronization effect
(Pfurtscheller et al, 1999)

Underlying idea

Taking advantage of a neurophysiological phenomenon to establish a communication between the brain and the computer



AN EXAMPLE OF BCI EXPERIMENT

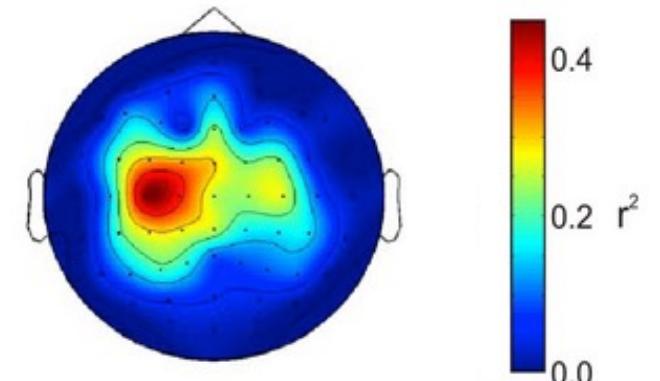


Behavioral properties

- Movement / preparation for movement : Event-related desynchronization (ERD) (Pfurtscheller, G, Lopes da Silva, FH, 1999)
- With relaxation/post-movement period : ERS

Why using it in BCI ?

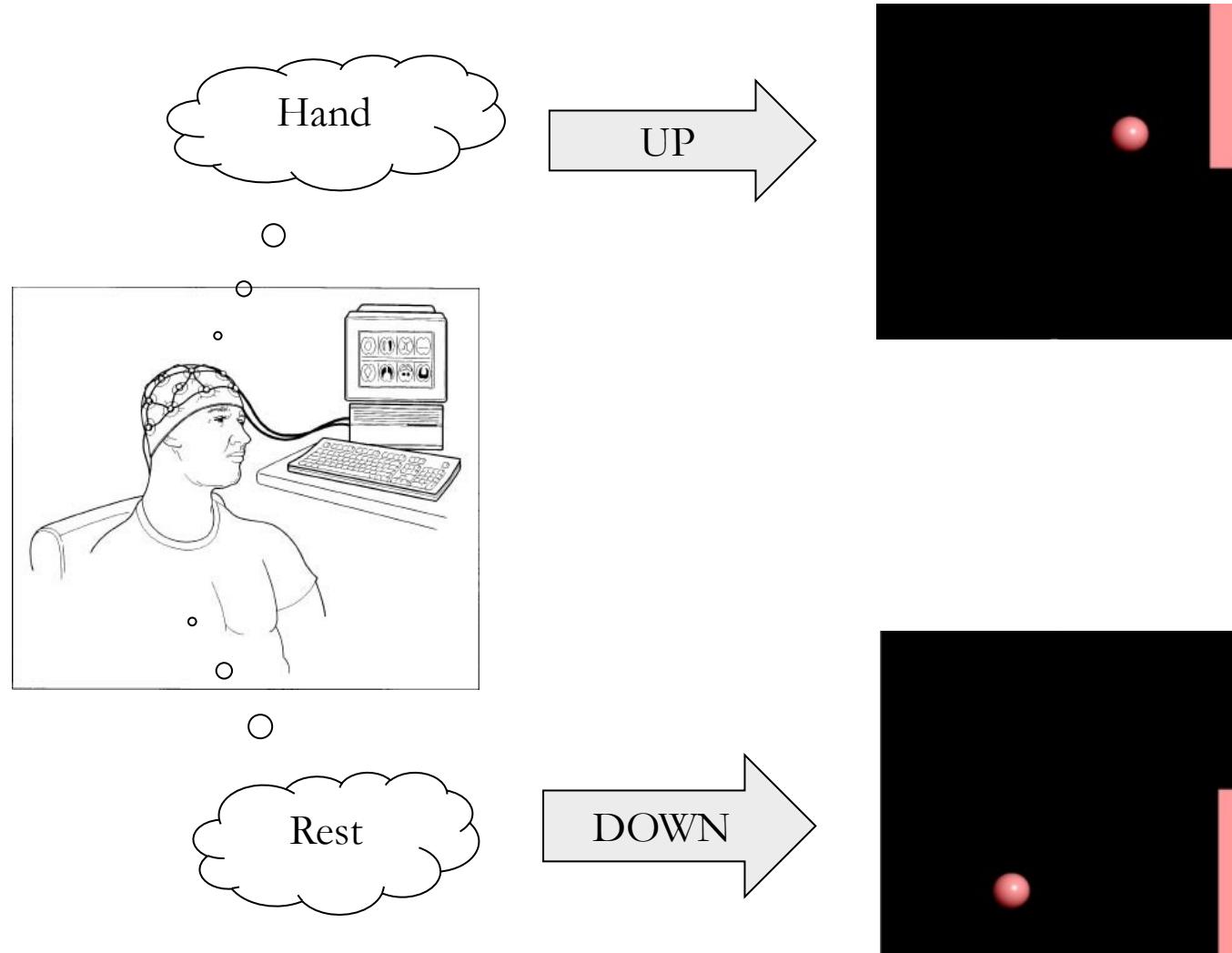
- Mu/Beta activity modulation by motor-imagery, a way to communicate
- Use of power spectra
- To establish this communication :
 - Spatial selection
 - Frequency selection



Illustrations from BCI2000 website

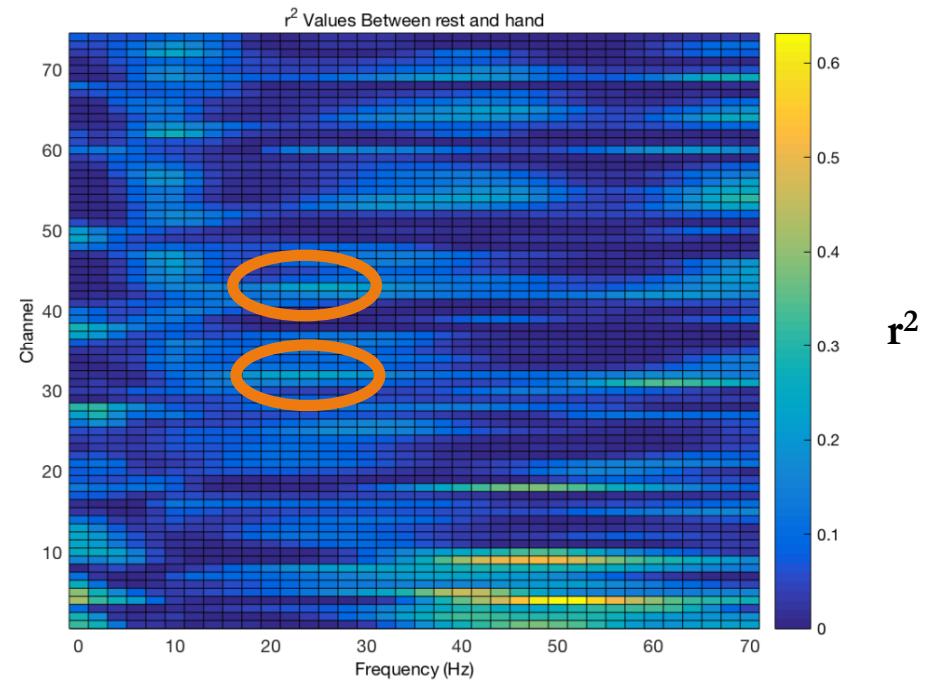
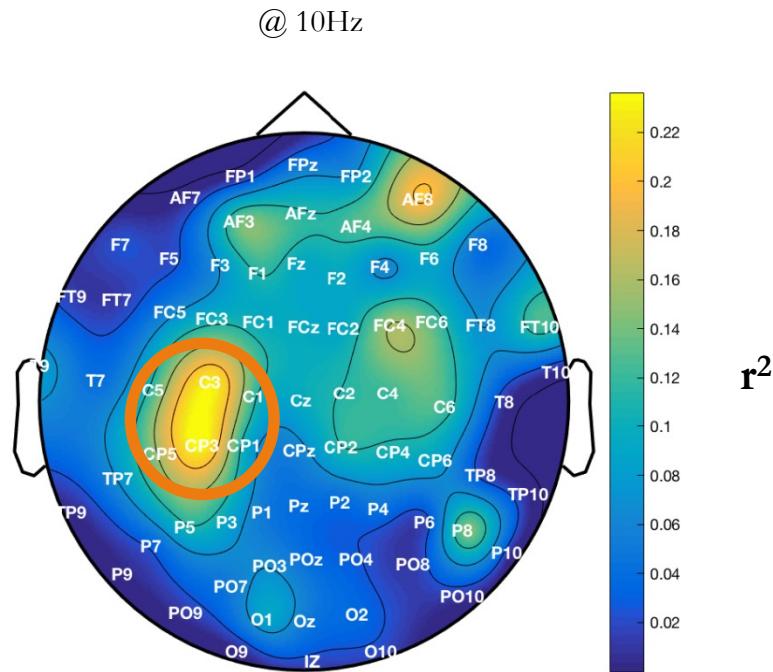
Motor imagery – In practice

21



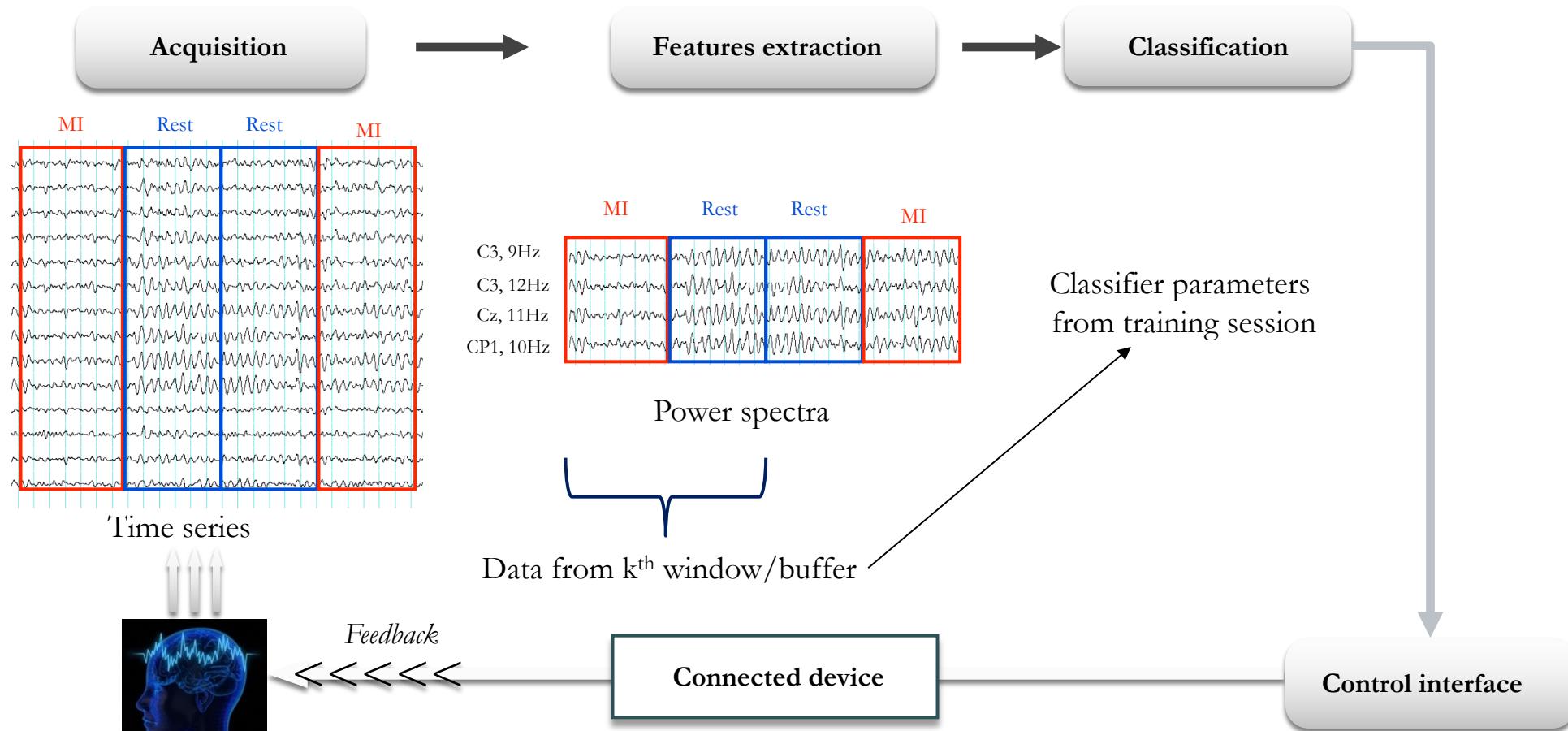
Motor imagery – In practice

22



Motor imagery – In practice

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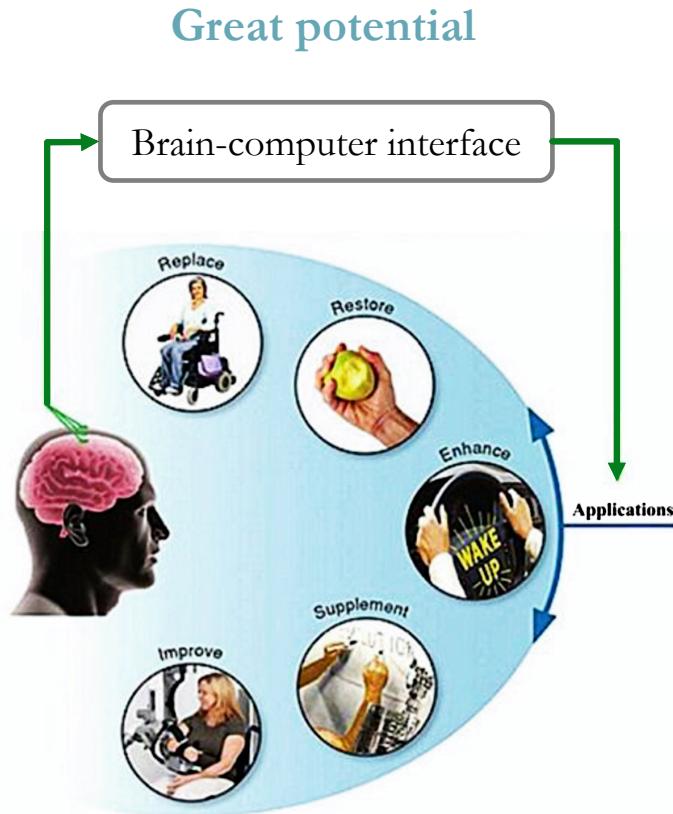


$$\text{cursor velocity} = \left(\sum_{i=1}^N \sum_{j=1}^{M=1} \text{Features}_{i,j} \text{Classifier}_{i,j,k} - \text{Offset} \right) \times \text{Gain}$$

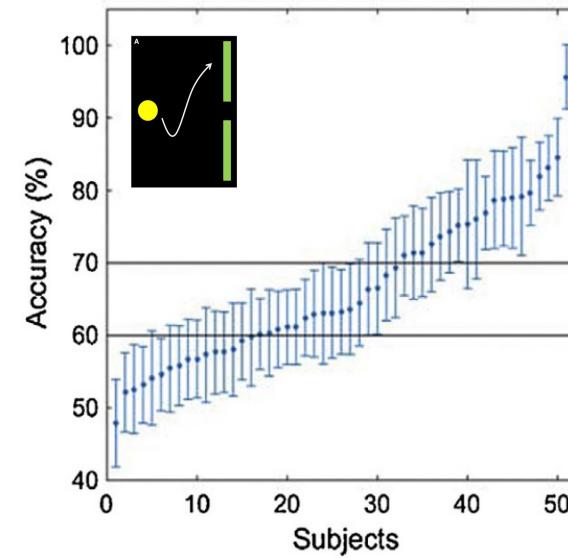
Motor imagery – In practice

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Poor usability



(Ahn & Jun, 2015)

Problem: Current BCIs fail to detect the mental intentions in ~30% of users –
BCI inefficiency (Thompson, 2018)

Controlling a BCI – A tale of two learners

Machine-centered approaches

- Signal processing (Vidaurre et al, 2011)
- Classification algorithms (Lotte et al, 2018)

⇒ Rely on EEG signals

User-centered approaches

- Neurophysiological patterns (Blankertz et al, 2010; Ahn et al, 2015)
- Human factors & Cognitive profile (Hammer et al, 2012; Jeunet et al, 2015)

⇒ Lack of reliable markers

Classification algorithms – classical methods

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Type	Examples of methods used in BCI
Linear classifiers	<ul style="list-style-type: none">- Linear Discriminant Analysis (LDA) and its regularized version- Support Vector Machine (SVM)
Neural networks (NN)	<ul style="list-style-type: none">- Multi-Layer Perceptron (MLP)
Non-linear Bayesian classifiers	<ul style="list-style-type: none">- Bayes quadratic classifiers- Hidden Markov Models (HMMs)
Nearest neighbour classifiers	<ul style="list-style-type: none">- k-Nearest Neighbour (kNN)- Mahalanobis distance classifiers
Classifier combinations	<ul style="list-style-type: none">- Boosting- Voting- Stacking

Current challenges:

- Low signal-to-noise ratio of EEG signals
- Non-stationarity over time of EEG signals
 - Inter/intra-subject variabilities
- Limited amount of training data for calibration
- Low reliability and performance of current BCIs

- Adaptive classifiers
- Classification of matrices
- Classification of tensors
- Transfer learning
- Deep learning

- Motivations
 - New subject/protocol – necessity to try out different experimental conditions
 - Working with patients means often short sessions to avoid tiredness and lack of motivation
 - Data change over time (non-stationary feature distributions) – cf rehabilitation training
- Methods
 - LDA variants with domain-specific regularizations [[Sosulski et al, 2021](#)]
 - Transfer learning from earlier sessions/other subjects/different tasks [[Kobler et al, 2022](#)]
 - Deep Learning and data augmentation [[Rommel et al, 2022](#)]
- Resources
 - Workshop dedicated to small datasets (BCI meeting 2023) by M. Tangermann – [materials](#) (slides and code)



CONCLUDING REMARKS

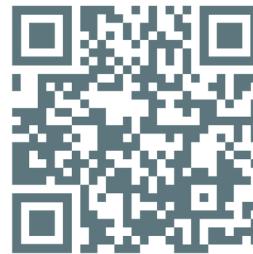
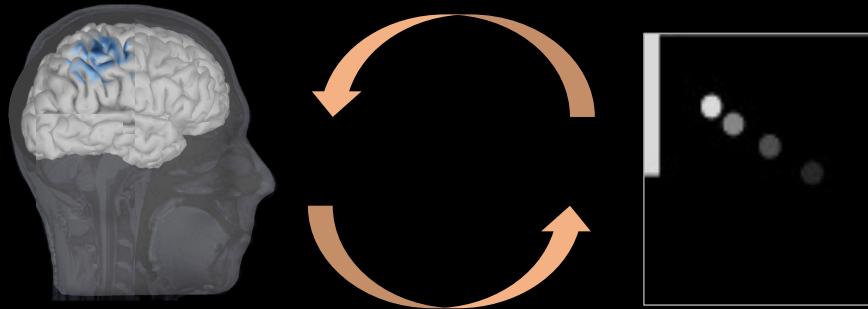
- BCI
 - Promising tool for clinical applications
 - **Multidisciplinary** domain
 - Growing interest in the last few years with the AI
- BCI learning & inter-subject variability
 - Improving the classifier / signal processing
 - Improving instructions
 - Finding (new) subject-related predictors
- Groups & events
 - International: [BCI society](#), international society
 - [Cybathlons](#): competitions to promote BCI and to test the finest algorithms with **end users !**
 - In France: [CORTICO](#), French association to promote BCI

- Tools – with many tutorials
 - Performing online experiments : [OpenViBE](#), an Inria software
 - Open datasets to test algorithms & check their replicability: [MOABB](#)
 - M/EEG data analysis : [MNE-Python](#)
 - Classification tools : [Scikit-learn](#)
 - Extracting and selecting features in BCI: [HappyFeat](#) an Inria software

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 - Extracting and selecting features in BCI: [HappyFeat](#) an Inria software
- M/EEG
 - [Chapter book](#)
 - Origins of the signals,
 - M/EEG experiments,
 - Data analysis,
 - Features extraction & selection,
 - Brain disorders

- Tools – with many tutorials
 - Performing online experiments : [OpenViBE](#), an Inria software
 - Open datasets to test algorithms & check their replicability: [MOABB](#)
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 - Extracting and selecting features in BCI: [HappyFeat](#) an Inria software
- M/EEG
 - [Chapter book](#)
 - [Github repo](#)
 - E/MEG data visualization,
 - Data extraction (ERD/S),
 - Classification

Thank you for your attention!



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arthur.desbois@inria.fr



MConstanceCorsi
ADesboisDev



[mccorsi](https://github.com/mccorsi)
[AsteroidShrub](https://github.com/AsteroidShrub)