





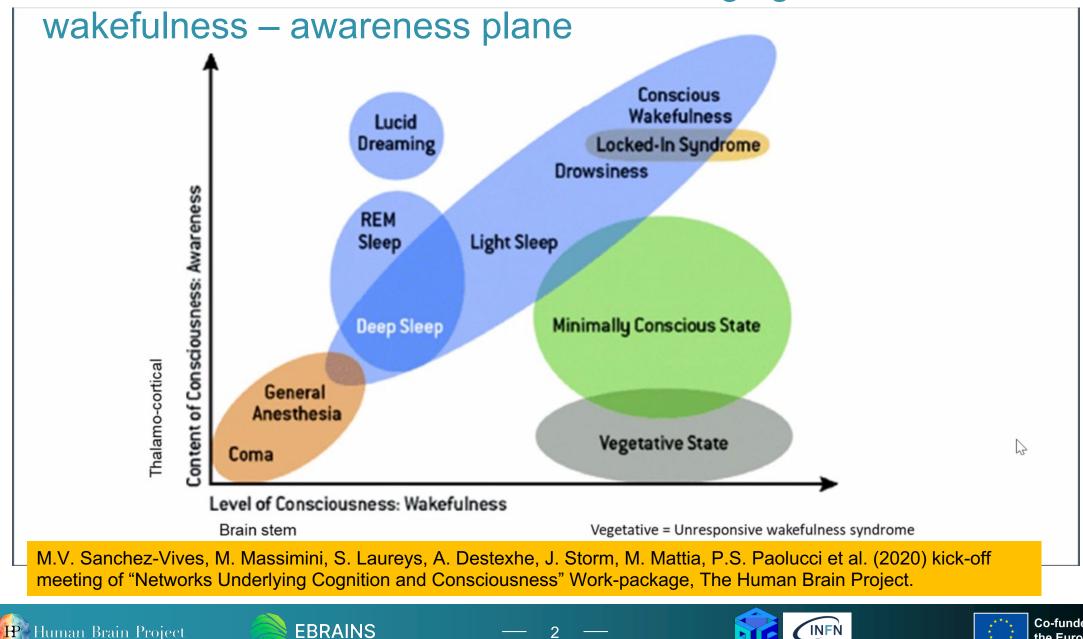
Cognitive and energetic benefits of awake/sleep cycles during incremental learning in spiking neural networks -> Bio-inspired AI

Pier Stanislao Paolucci

On behalf of APE Lab @ INFN: R. Ammendola, I. Bernava, A. Biagioni, G. De Bonis, C. Capone, P. Cretaro, C. De Luca, O. Frezza, F. Lo Cicero, A. Lonardo, C. Lupo, M. Martinelli, P.S. Paolucci, E. Pastorelli, L. Pontisso, F. Simula, L. Tonielli, M. Turisini, P.Vicini



Brain States and Cosciousness: an emerging classification in the



Co-funded by the European Union

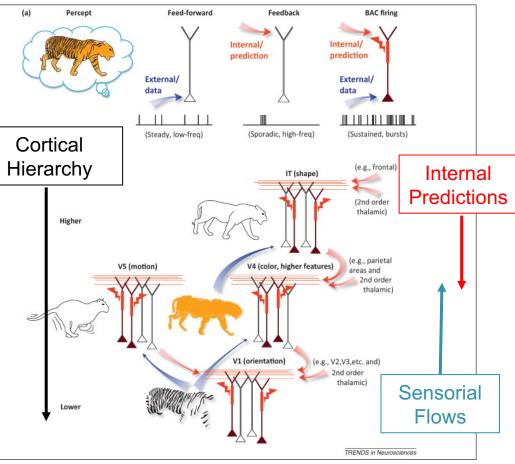
Thalamo-cortical spiking models showing the beneficial cognitive and energetic effects of the interplay among sleep and memories, learned by combining contextual and perceptual information

- Sleep essential, in all animal species
- Young humans pass the majority of time sleeping, when learning is faster
- Sleep deprivation detrimental for cognition, even in adults
 - Sleep deprivation, terrible torture
- Roles in biological intelligence
 - Optimization of energy consumption
 - Homeostatic processes (normalization of representations)
 - Novel, creative associations and planning
 - Optimization of performances
 - Recovery /restorations of bio-chemical optimality
- (our opinion) Sleep essential for bio-inspired artificial intelligence

Thalamo-cortical spiking model of incremental learning combining perception, context and NREM-sleep **PLoS Computational Biology** (2021). B.Golosio, C. De Luca, C. Capone, ..., P.S. Paolucci. <u>https://doi.org/10.1371/journal.pcbi.1009045</u>

Sleep-like slow oscillations improve visual classification through synaptic homeostasis and memory association in a thalamo-cortical model *Scientific Reports* (2019). *C. Capone, E. Pastorelli, B. Golosio, P.S. Paolucci.* https://www.nature.com/articles/s41598-019-45525-0





Larkum, M. A cellular mechanism for cortical associations: an organizing principle for the cerebral cortex. *Trends in Neurosciences,* 36 (2013), 141.

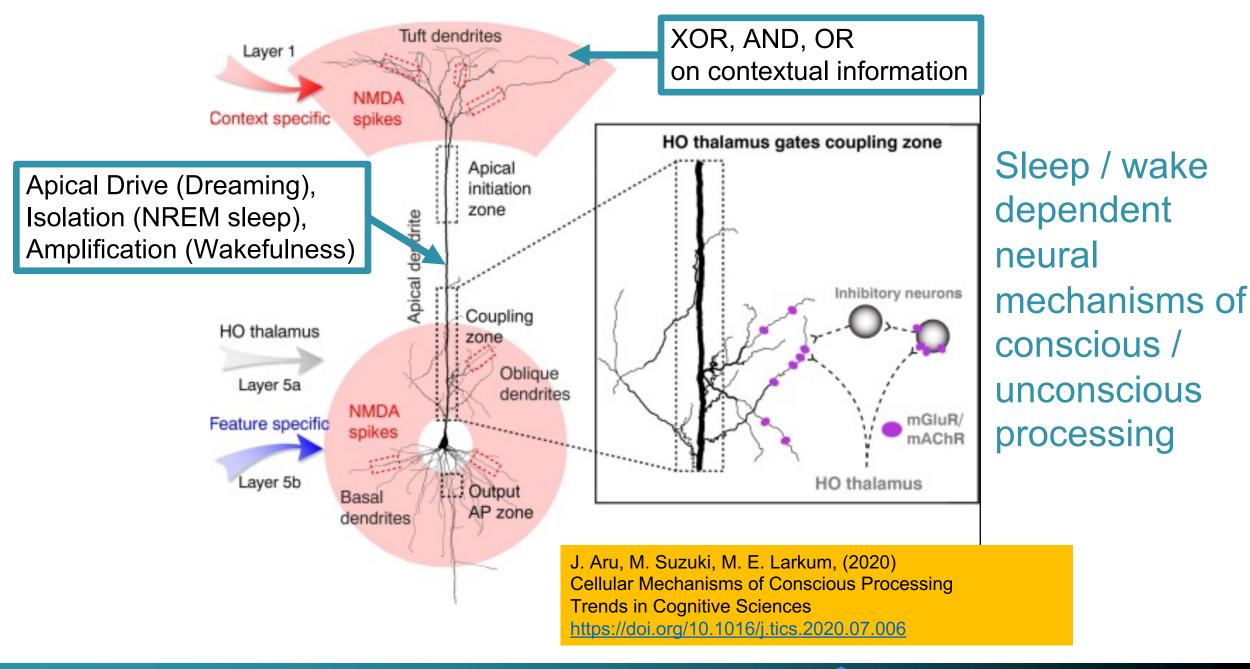




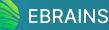
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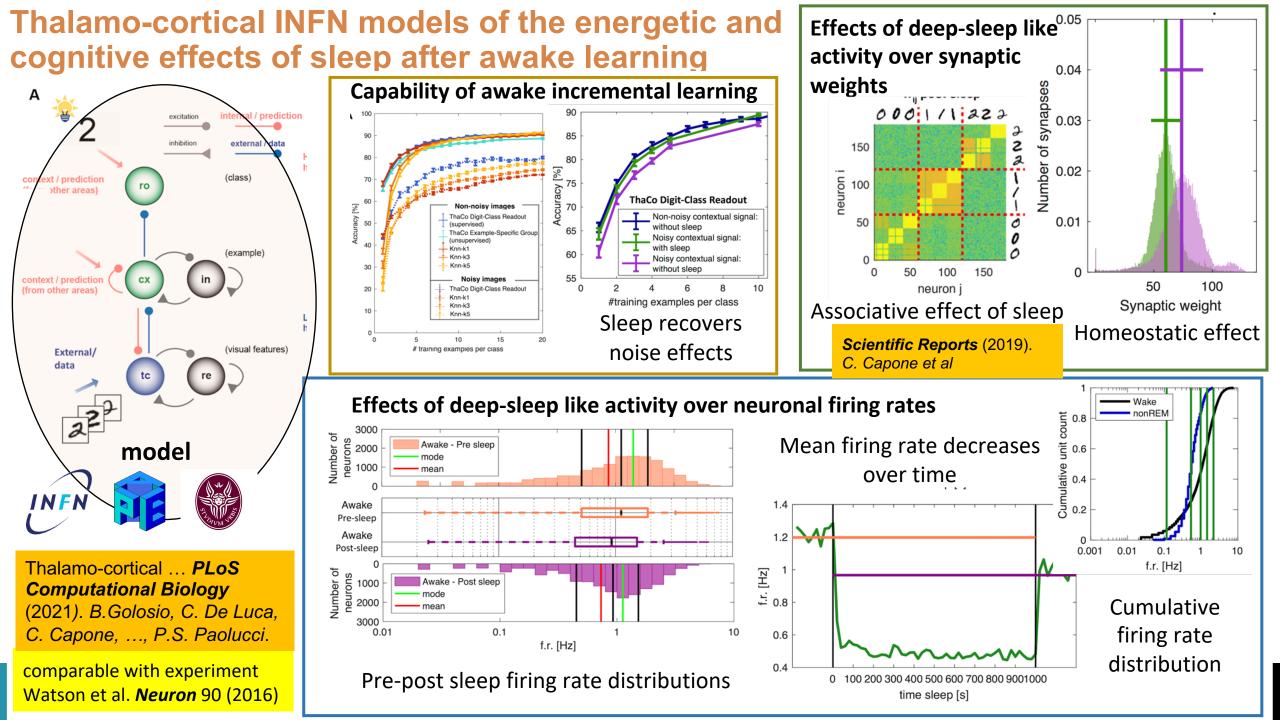












In progress INFN work to add dreaming (REM) to deep-sleep (NREM) and wakefulness modeling AWAKE NREM REM Apical amplification Apical isolation Apical drive

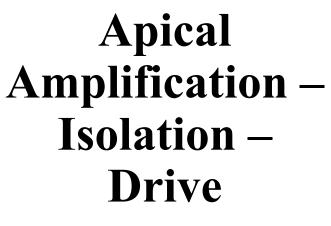
Cellular mechanism supporting "brain state specific" combination/decoupling of contextual information (i.e. internal information, flowing either Top-Down from areas higher in the abstraction Hierarchy and/or Laterally from other areas at similar level of abstraction With

bottom-up evidence from sensory system or areas lower in the abstraction hierarchy

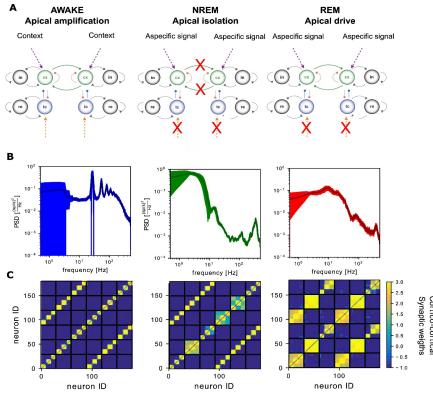
J. Aru, F. Siclari, W. A. Phillips, J. F. Storm (2020) Apical drive—A cellular mechanism of dreaming? Neuroscience & Biobehavioral Reviews

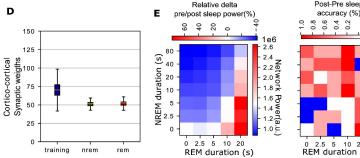
J. Aru, M. Suzuki, M. E. Larkum, (2020) **Cellular Mechanisms of Conscious Processing** Trends in Cognitive Sciences

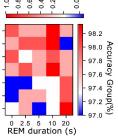




VS Wakefulness, NREM, REM











EBRAINS

INFN APE Lab in Human Brain Project - EBRAINS

HBP (2014-2023). budget > 500M€. (<u>https://www.humanbrainproject.eu</u>) >120 research institutions

- Mission: understand structure and mechanisms supporting brain functions: since ever a major philosophical challenge. Turned by experimental tools and HPC simulation in a scientific challenge
 - Physicists and HPC scientists central: complex systems, massive data analysis, models on HPC
- Societal motivations:
 - 800 B€/year in Europe, cost of neurologic traumatic and clinic diseases
 - Next-generation Bio-inspired AI
- Deliver of first release EBRAINS Research Infrastructure. (<u>https://ebrains.eu/</u>)
 - Open access / reproducible brain research. FAIR methodology. Integration of experimental data (atlases and activity) and models. (2023-...) EBRAINS European ESFRI: central structure + national nodes

APELab - INFN Roma: 2.1 M€ from HBP (2016-2023), "Networks underlying cognition and consciousness"

- Modeling cognitive, entropic and energetic optimizations effects produced by wakefulness, deep sleep (NREM), dreaming (REM) + high resolution cortical modelling on parallel/distributed computing systems
- Analysis of experimental data and high-resolution modeling of spatio-temporal features of brain activity waves
- Synergy with APELab "traditional" HPC projects. Benchmarks for next computing and networking architectures
- -> PNRR: EBRAINS Italian Node (2023-2026) proposal submitted Feb 2022 + Bio-Inspired AI (TBC)











BASSES Workshop

P Human Brain Project









EBRAINS Workshop

Brain Activity across Scales and Species:

Analysis of Experiments and Simulations (BASSES)

13 - 15 June 2022 **ROME (ITALY) AND ONLINE**

> Abstract Submission Deadline 6 MAY 2022 **Registration Deadline**

27 MAY 2022

Confirmed speakers

Jan Biaalie (University of Oslo)

The goal of the BASSES Workshop is to provide an overview of the scientific topics of brain states and complexity, state transitions, and their connection with cognitive functions, and to demonstrate the achievements in this field obtained within the Human Brain Project thanks to the functionalities provided by the EBRAINS research platforms.

BASSES will allow people with different expertises, from experimental and theoretical neuroscientists to computer scientists, to share results and ideas and connect into a wider community.

Scientific chairs

Anna Letizia Allegra Mascaro | LENS Giulia De Bonis | INFN

Hands-on sessions

Handling EBRAINS data Lyuba Zehl | Forschungszentrum Jülich

Running analysis in EBRAINS Michael Denker | Forschungszentrum Jülich

Simulating spatially organised networks with NEST Johanna Senk | Forschungszentrum Jülich

Validating models against data in EBRAINS Andrew Davison | CNRS

Alessandra Camassa (IDIBAPS) Cristiano Capone (INFN) Chiara De Luca (INFN) Alain Destexhe (CNRS) Jennifer Goldman (INFN) Bruno Golosio (University of Cagliari Robin Gutzen (Forschungszentrum Jülich) Viktor Jirsa (Aix-Marseille University) Arnau Manasanch (IDIBAPS) Thierry Nieus (University of Milan Marcello Massimini (University of Milan) Maurizio Mattia (ISS) Elena Montagni (LENS) Pier Stanislao Paolucci (INFN) Francesco Pavone (LENS) Elena Pastorelli (INFN) Andrea Pigorini (University of Milan Mavi Sanchez-Vives (IDIBAPS)

Johan Storm (Unversity of Oslo)

Sacha van Albada (Forschungszentrum Jülich)

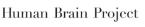
Gianni Valerio Vinci (ISS)

Further information

www.humanbrainproject.eu/en/education/ebrains-workshops/basses



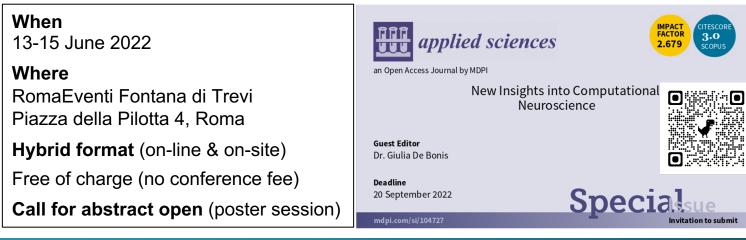






Organized by APELab/LENS on behalf of the Human **Brain Project – EBRAINS Research Infrastructure**

BASSES:= Brain Activity across Scales and Species: Analysis of Experiments and Simulations

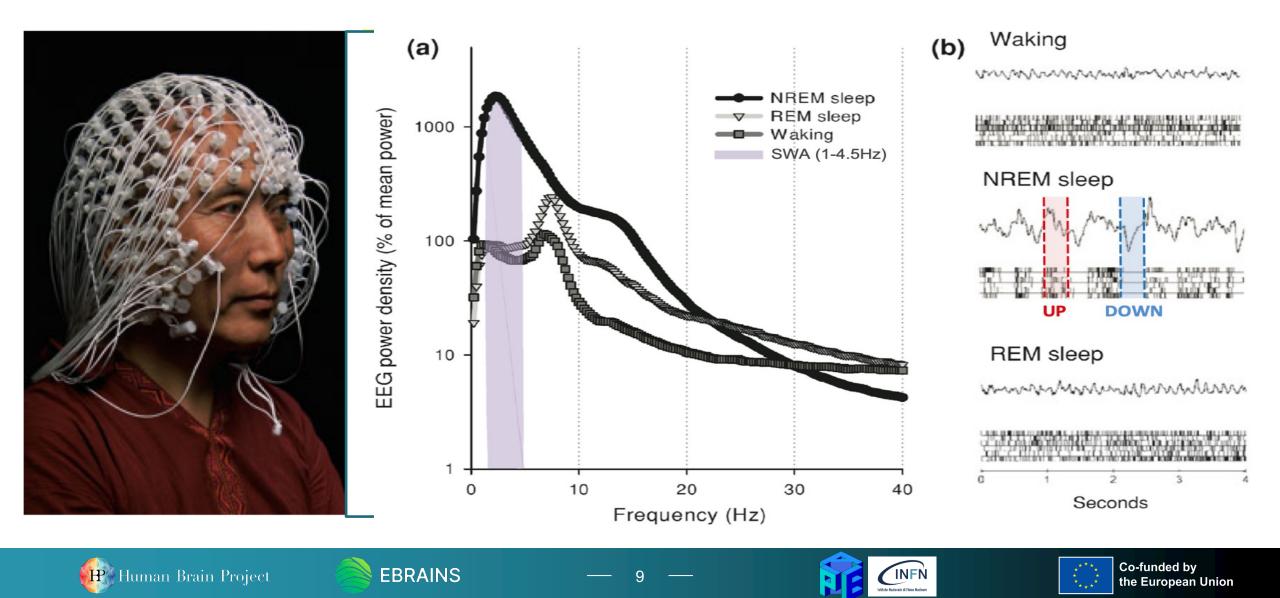


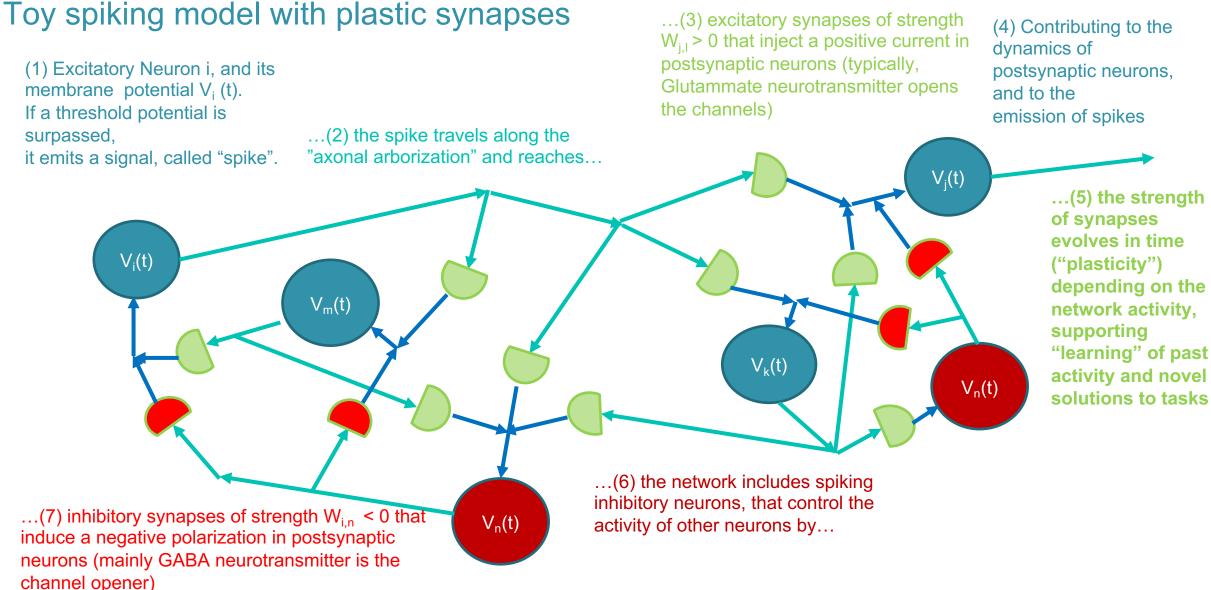
EBRAINS





Deep-sleep NonREM <-> Unconscious state, Slow Waves in Delta band REM (dreaming) and Wake <-> Conscious, Asyncronous Irregular activity





P Human Brain Project







Toy model: networks of neurons with Spike Frequency Adaptation can enter and exit from the NREM Slow Wave Activity regime

V(t), membrane potential of the neuron C_m , the membrane capacitance of the neuron

 $\omega,$ the neural *fatigue* due to the emission of spikes

 I_{system} , the input current, spikes from other neurons injected by synapses

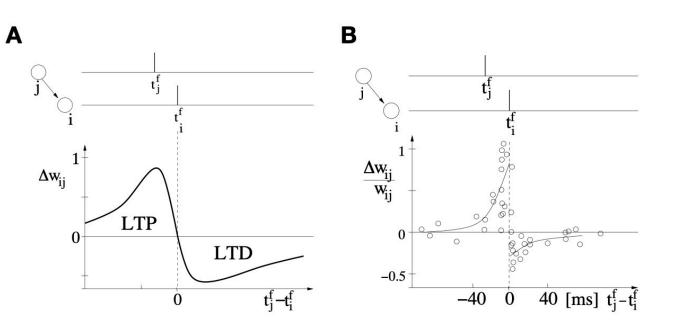
$$\begin{cases} V(t) \geq V_{treshold} \rightarrow & \text{spike emission, axonal transmission to} \\ & \text{synapses that injects current in other neurons} \\ & \text{reset potential to afterspike } V_{afterspike} \\ & \text{increase the fatigue } \omega \\ \\ C_m \frac{dV}{dt} & = -g_L \left(V - E_L\right) + g_L \Delta_T e^{\frac{\left(V - V_{th}\right)}{\Delta_T}} + I_{system} - \omega \\ \\ & \tau_\omega \frac{d\omega}{dt} & = a \left(V - E_L\right) + b \sum_k \delta(t - t_k) - \omega \end{cases}$$
(1)





Change b parameter

Learning in a toy spiking model: Spike Timining Dependent Plasticity (STDP) • Synapses are engines th



General form of synaptic Spiking Time Dependent synaptic Plasticity (STDP)

$$\Delta w(\Delta t) = \begin{cases} A_+ e^{\frac{\Delta t}{\tau_+}} & \Delta t > 0\\ A_- e^{\frac{\Delta t}{\tau_-}} & \Delta t \le 0 \end{cases}$$

- Synapses are engines that are able to detect both causality and anti-causality.
- They can reward themselves, by increasing their values when causality is detected,
- and depress themselves when anti-causality is detected.

Specific form used in many of our simulations: NLTAH-STDP

$$\Delta w(\Delta t) = \begin{cases} -\lambda \alpha w^{\mu} e^{-\frac{\Delta t}{\tau}} & \Delta t \le 0\\ \lambda (1-w)^{\mu} e^{-\frac{\Delta t}{\tau}} & \Delta t > 0 \end{cases}$$





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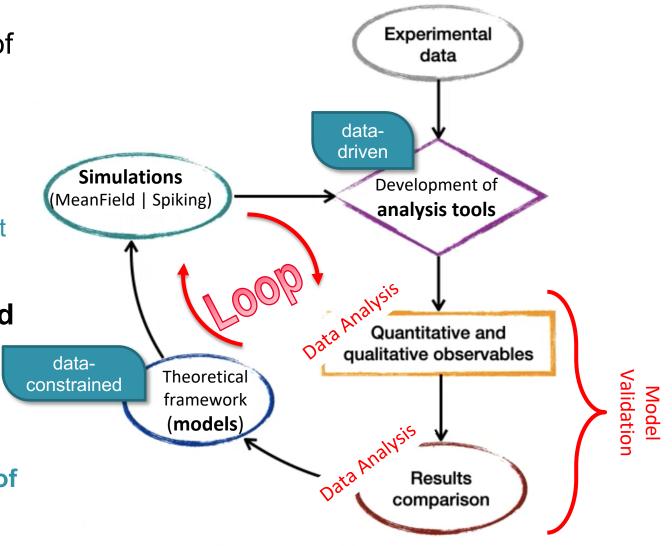
Interplay DataAnalysis – Models – Simulations

understand mechanisms and features of the brain dynamics from **observation** and **interpretation** of **experimental data**

 define benchmark observables and design data-driven analysis tools, for comparing and combining different datasets, aiming at general claims and at statistically significant assessments

extract results from experimental recordings for feeding data-constrained simulations and refining theoretical models

 define methods and procedures for the validation of theoretical models (comparison of experimental and simulated data) and for the comparison of models







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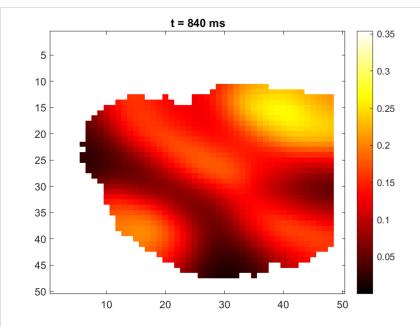




INFN: large scale simulations

- Spiking and mean field neural network simulations reproducing models of cortical slow wave activity at the scale of mouse hemisphere
- Data-driven model: simulation parameters inferred from experimental data (wide-field calcium imaging)





- Near to biological resolution (54K neuron/mm2 and about 5K synapses per neuron in rat neocortex area)
- Simulated and experimental behavior compared using analysis tools
- Model used as a benchmark to drive the development of future interconnects for platforms including millions of embedded ARM cores

















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