



Recent Advances in AI and Machine Learning @ OsloMet AI Lab

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URL: <https://ailab.oslomet.no/>

A Snapshot of OsloMet AI Lab

MISSIONS

- conduct **cutting-edge AI research**
- **educate** graduate (MS and PhD) students and train postdocs.
- **bridge the gap** between AI theoretical research and real-world applications.
- host **academic conferences** and other scientific events.
- foster close **partnership** with national and international **research groups and companies** in the AI field.



Home

The **OsloMet Artificial Intelligence Lab** is a joint [OsloMet](#) and [SimulaMet](#) research centre in the heart of Oslo. The OsloMet AI Lab is hosted by the [Department of Computer Science](#), located in Pilestredet 52 (Oslo). OsloMet AI Lab was initiated by the [Applied AI research group](#) and now includes members from many research groups at OsloMet and SimulaMet. For a list of groups and members see [here](#).

The OsloMet AI lab conducts both applied and fundamental research projects and student projects in artificial intelligence, including theory and applications of machine learning in several domains.

The OsloMet AI Lab positions OsloMet's AI research environment at the forefront of AI in Norway, will educate students within AI and attract talents.

The OsloMet AI Lab involves more than 40 Professors and Associate Professors from different research groups, 5 Senior Researchers, and 20 PhDs and Post-Docs, in addition to several bachelor and master students.

Stay tuned!

OsloMet AI Lab Board



[Hugo Lewi Hammer](#)
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Coordinator



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Members

Different **research groups**:

- [Applied Artificial Intelligence](#), IT, OsloMet
- [Machine Intelligence Department](#), SimulaMet (Simula Metropolitan Center for Digital Engineering)
- [Living Technology Lab](#)
- [Autonomous Systems and Networks](#)
- [Mathematical Modelling](#)
- [Automation, Robotics and Intelligent Systems](#), MEK, OsloMet
- [Motion Analysis Lab](#)
- [Digital Innovation and Strategic Competence in Organizations](#)

Figures:

- > 40 Professors and Associate Professors
- 5 Senior Researchers
- 20 PhDs and Postdocs

Focus R&D Areas

FUNDAMENTAL:

- **Machine learning algorithms and optimization:** Improvements of **ML and other AI models** and hard **multi-objective optimization** problems (Supervised, unsupervised, and reinforcement learning; Deep/wide learning; Neuro-evolution; Multi-objective evolutionary algorithms)
- **Computational intelligence:** **Recurrent neural networks; Fuzzy systems; Evolutionary computation**
- **Complex systems and artificial life**

APPLIED:

- **Robotics and IoT:** **Adaptive and Autonomous Systems** in robotics (HW - vehicles/drones; SW – intelligent control, knowledge-based systems) and IoT across various application domains.
- **Healthcare:** Using AI as **intelligent decision support** or to develop **personalized/individualized systems** based on learning from the users.
- **Neuroengineering:** **Neurocomputing; Neuroergonomics; Brain-machine interaction;** Human-machine symbiosis; Adaptive automation

PARTNERS

- [IBM Norge](#)
- [Norwegian Artificial Intelligence Research Consortium \(NORA\)](#)
- [Confederation of Laboratories for Artificial Intelligence Research in Europe \(CLAIRE\)](#)
- [Norwegian Open AI Lab](#)

ONGOING PROJECTS (SELECTED)

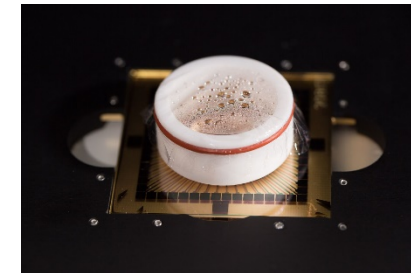
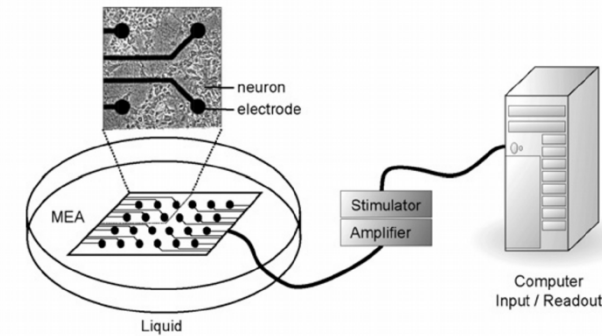
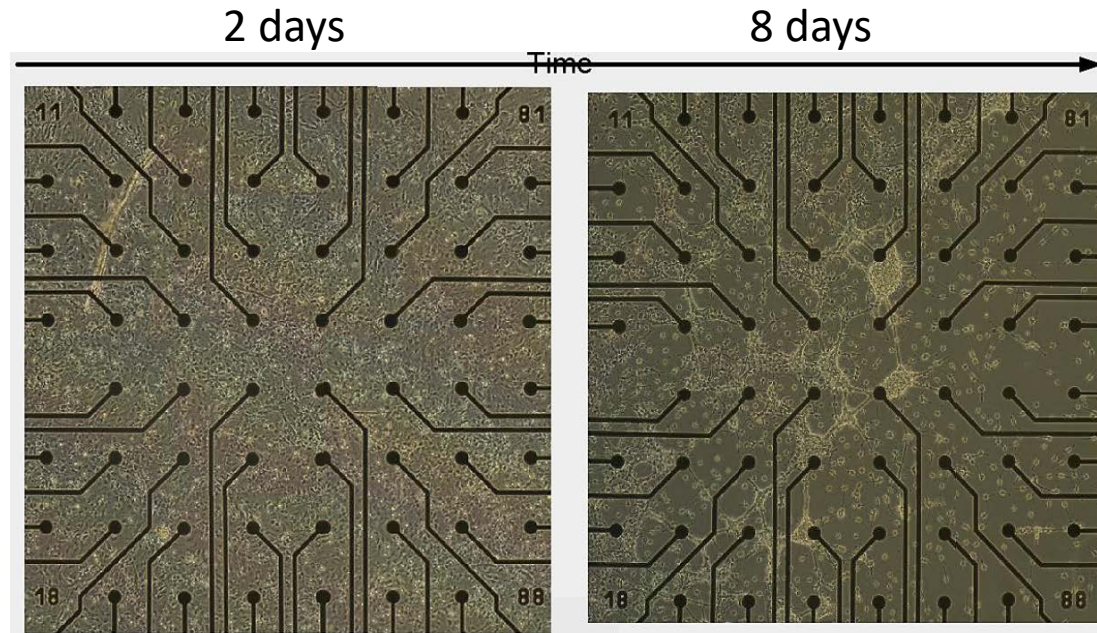
Third-party funding:

- **Socrates** – Self Organising Computational Substrates, **NFR**, Nichele
- **OASYS** – Ocean-Air synoptic operations using coordinated autonomous robotic SYStems and micro underwater gliders, **EU**, Alcocer
- SCOTT – Secure Connected Trustable Things, **EU**, Van Do & Feng
- Pacer – Patient Centric Engineering in Rehabilitation, **NFR**, Mirtaheri & Hammer & Yazidi
- Artificial intelligence – a novel tool in assisted reproduction technology, **NFR**, Hammer & Yazidi
- DeepCA – Hybrid Deep Learning Cellular Automata Reservoirs, **NFR**, Nichele

Internally funded:

- **Adaptive Automation** of Safety-Critical Human-Machine Systems, Zhang
- **FELT** – Futures of Living Technologies, Bergaust
- CAOS – Complex Adaptive and Self Organising Systems, Nichele
- Personalised Cervical Cancer from Historical data, Naumova
- Multi-modal Data Fusion based on Coupled Factorizations, Acar Ataman

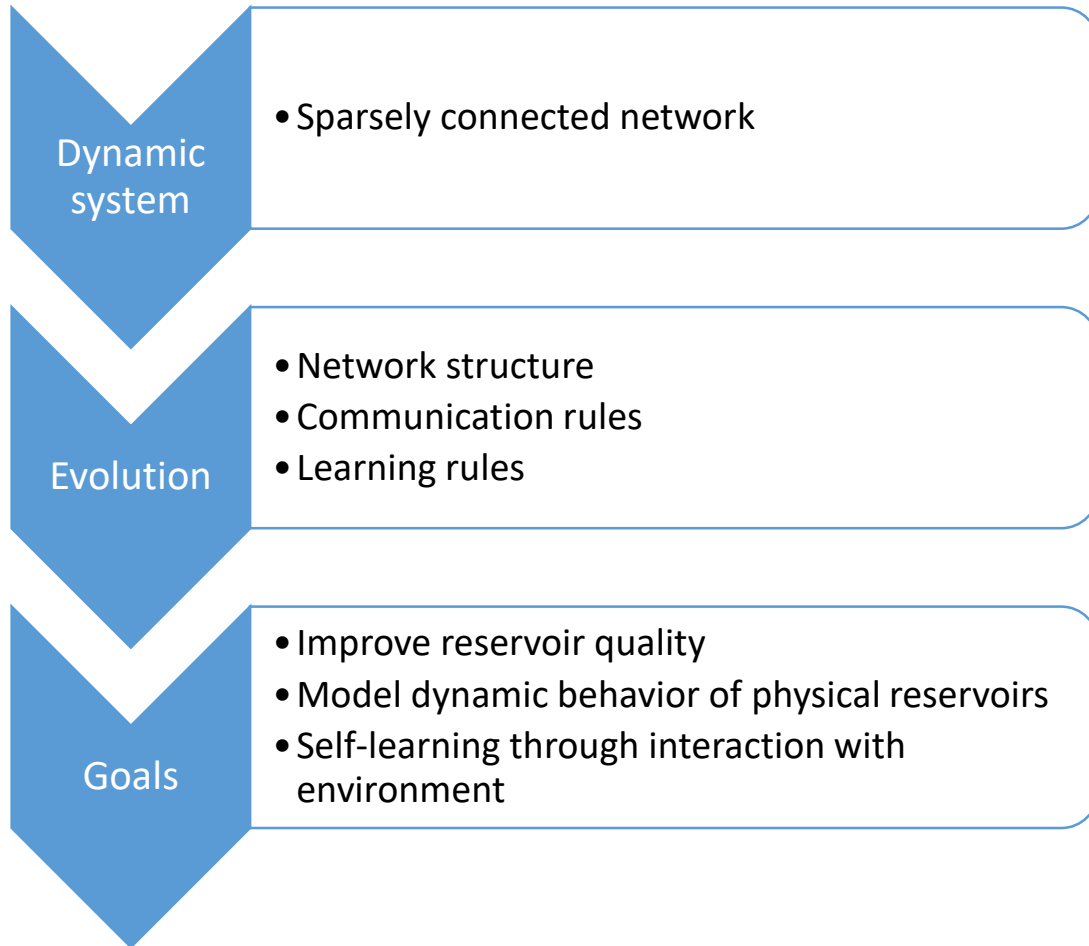
TOPIC 1: NFR Socrates - Biological Neural Networks



- Better understanding of neural networks
- Medical: right neural stimulations for rehabilitation/treatment

P. Aaser, M. Knudsen, O. Ramstad, R. van de Wijdeven, S. Nichele, et al. (2017), [Towards Making a Cyborg: A Closed-Loop Reservoir-Neuro System](#), *Proc. of European Conf. on Artificial Life (ECAL) 2017*. (BEST POSTER AWARD)

EvoDynamic: EVOLution of discrete DYNAMIC systems based on self-organization through local interactions



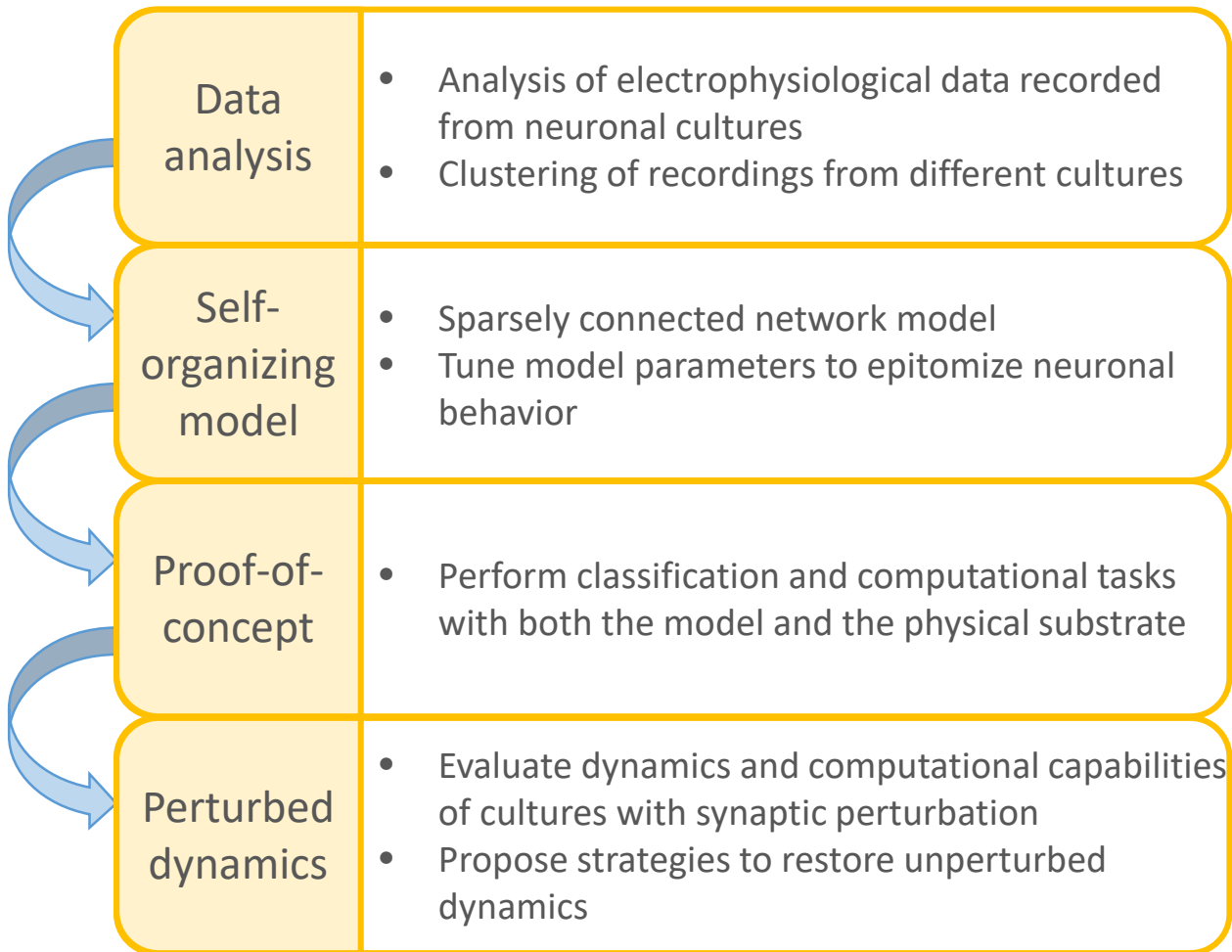
- Sparsely connected network for reservoir computing
- Dynamics of reservoir simplify complex inputs

[Schrauwen et al., ESANN, 2007]

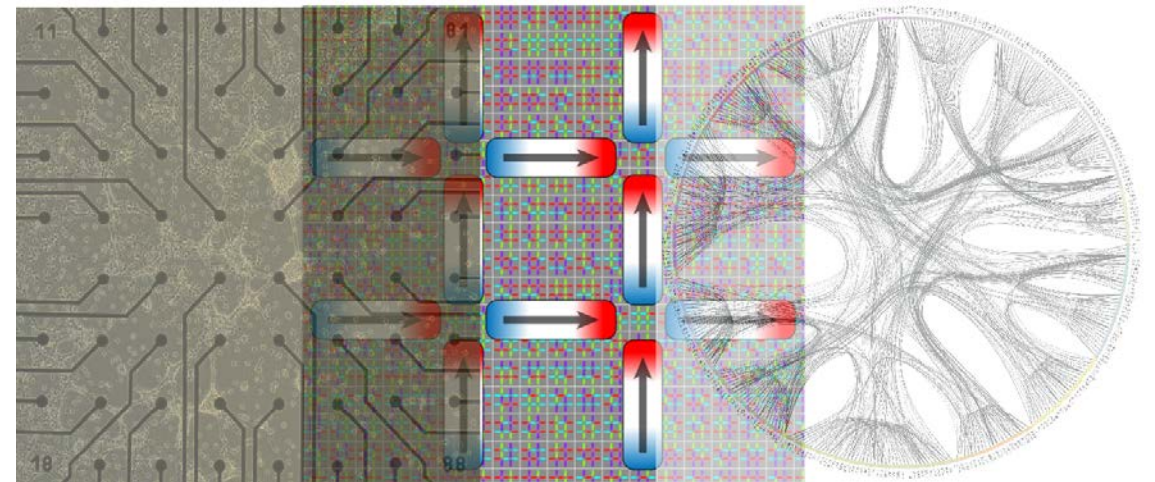
- Physical reservoirs for SOCRATES project¹
 - Biological neurons over microelectrode arrays
 - Nanomagnetic ensembles

¹ <https://www.ntnu.edu/socrates>

SOMA – Self-Organizing Models of Artificial learning in neural substrates



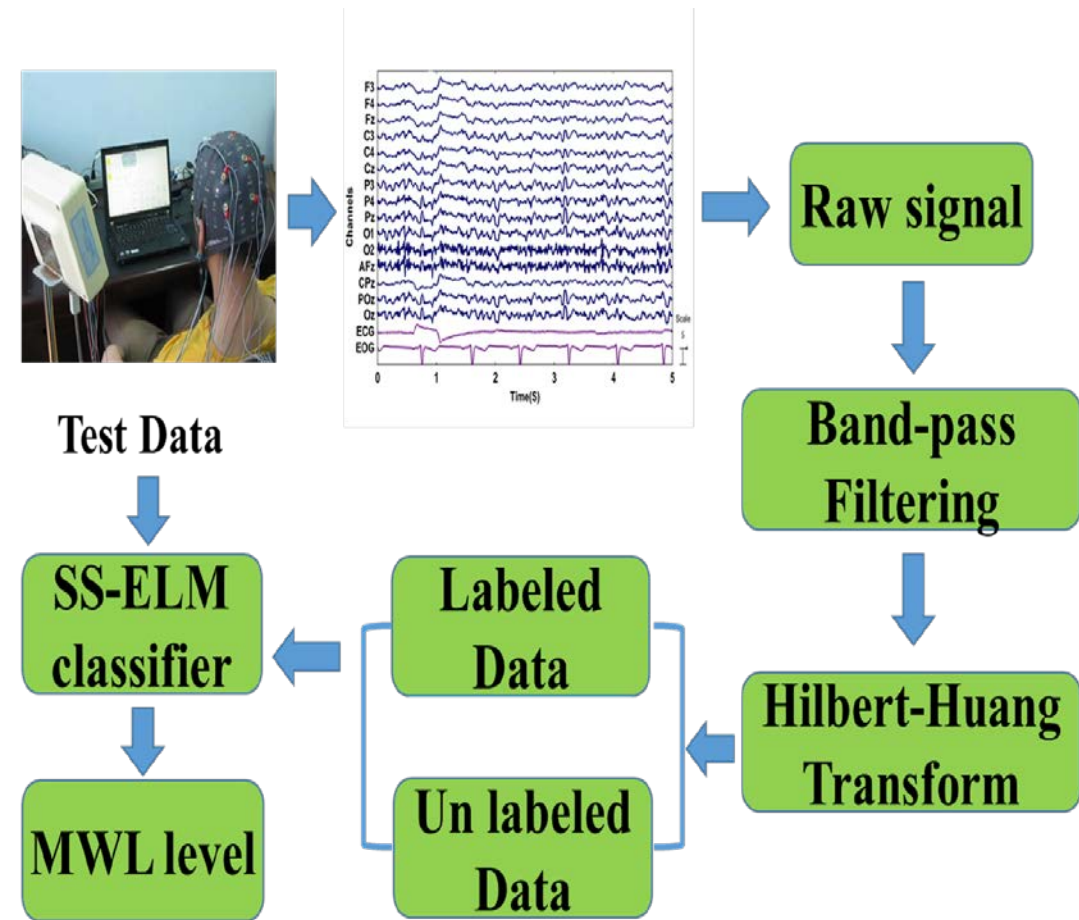
SOCRATES project:
Inspired by the self-organizing behavior of neurons, to develop arrays of nanomagnets for new computing hardware that is scalable, energy efficient, fault tolerant, and self-learning.
<https://www.ntnu.edu/socrates>



Kristine Heiney, PhD fellow
Supervisor: Stefano Nichele

TOPIC 2: Pattern Classification of Cognitive Workload based on Semi-Supervised Learning

- The real-time mental workload (MWL) monitoring is crucial for designing **adaptive aiding/assistance** systems.
- Although data-driven approaches have potential for MWL recognition, it is usually difficult or expensive to acquire **sufficient labeled data**.
- This work applied **semi-supervised extreme learning machine** to MWL classification based only on a **small number of labeled physiological data**.



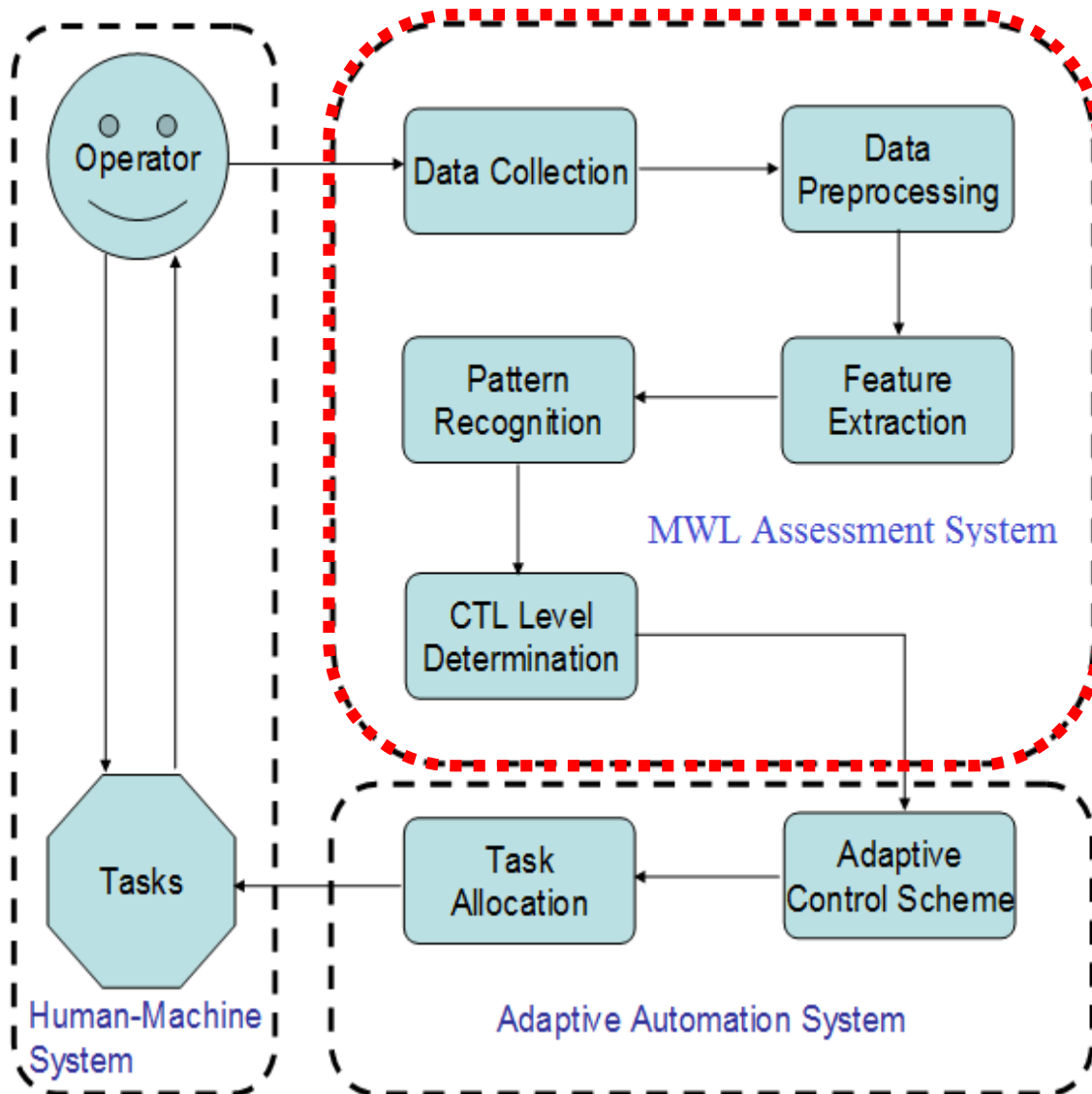
Conclusion

- The proposed SS-ELM method can effectively improve the accuracy and efficiency of MWL classification.
- When only a **small number of labeled data** are available in practice, SSL algorithm is suitable for **online real-time** MWL pattern recognition task.

References

1. **J. Zhang***, X. Cui, J Li and R. Wang, Imbalanced Classification of Mental Workload Using a Cost-Sensitive Majority Weighted Minority Oversampling Strategy, *Cognition, Technology & Work*, vol. **19** (4), pp. 633-653, 2017.
2. J. Li, **J. Zhang***, J. Xia and P. Chen, Mental workload classification based on semi-supervised extreme learning machine, *J. of East China University of Science and Technology* (Natural Sci. Ed.), 2019 (to appear).
3. J. Li and **J Zhang**, Mental workload classification based on semi-supervised extreme learning machine, *26th Int. Conf. on Artificial Neural Networks (ICANN17)*, 11-15 Sep. 2017, Alghero, Sardinia, Italy.

TOPIC 3: Pattern Classification of Cognitive Workload Using Deep Learning



Motivations

Mental Workload (MWL) is an important indicator of mental activity of human operator in Human-Machine System (HMS).

Aims & Objectives

- ✓ Recognition of Momentary MWL using electrophysiological data.
- ✓ Application of deep learning to MWL classification.

Conclusion

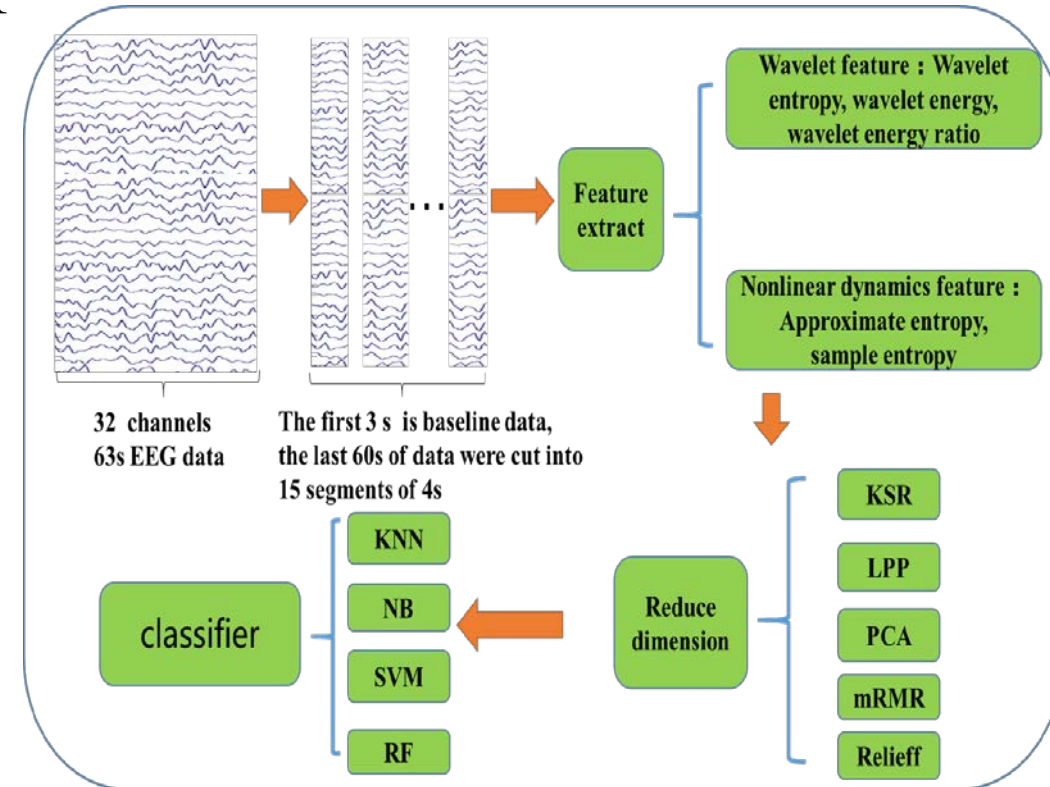
- The two deep learning architectures proposed can extract features automatically and are also computationally efficient.
- The selected optimal EEG channels can be used to design wearable devices for high-risk MWL detection.

References

1. **J. Zhang***, S. Li and R. Wang, Pattern Recognition of Momentary Mental Workload Based on Multi-Channel Electrophysiological Data and Ensemble Convolutional Neural Networks, *Front. Neurosci.*, vol. **11**, Article 310, pp. 1-16, May 30, 2017, doi: 10.3389/fnins.2017.00310.
2. **J. Zhang***, Y. Wang, and S. Li, Cross-subject Mental Workload Classification Using Kernel Spectral Regression and Transfer Learning Techniques, *Cognition, Technology & Work*, vol. **19** (4), pp. 587-605, 2017.
3. **J. Zhang*** and S. Li, A Deep Learning Scheme for Mental Workload Classification based on Restricted Boltzmann Machine, *Cognition, Technology & Work*, vol. **19** (4), pp. 607-631, 2017.
4. J. Li and **J Zhang**, Mental workload classification based on semi-supervised extreme learning machine, in A. Lintas, S. Rovetta, P. Verschure and A. Villa (Eds.), *ICANN 2017, Part II, LNCS 10614*, pp. 297-304, Springer Int. Publ. AG, 2017.
5. **J. Zhang***, S. Li and R. Wang, Pattern Classification of Instantaneous Mental Workload Using Ensemble of Convolutional Neural Networks, in *Proc of 20th IFAC World Congress*, Toulouse, France, July 2017.

TOPIC 4: EEG-based Emotion Recognition Using Machine Learning

- Used clustering to determine **4 target classes** of human emotion.
- Performance comparisons:
 - **2 feature extraction** methods: wavelet transform, nonlinear dynamics
 - **5 feature reduction** algorithms: KSR, LPP, mRMR, ReliefF, PCA
 - **4 classifiers**: k-nearest neighbor (kNN), naïve Bayesian (NB), support vector machine (SVM), random forest (RF)



Conclusion

- 4-class emotion classification accuracy can be significantly improved by taking into account baseline EEG features.
- Nonlinear dynamics features lead to higher accuracy than wavelet-derived features.
- EEG gamma-band features are more salient than other frequency bands.
- Best combination: Kernel Spectral Regression (KSR) for dimensionality reduction + RF for classification.

References

1. P. Chen and J Zhang, Performance comparison of machine learning algorithms for EEG-signal-based emotion recognition, in A. Lintas, S. Rovetta, P. Verschure and A. Villa (Eds.), *ICANN2017, Part I, LNCS 10613*, pp. 208-216, Springer Int. Publ. AG, 2017.
2. P. Chen, J. Zhang*, Z. Wen, J. Xia and J. Li, EEG-based emotion recognition through kernel spectral regression and random forest approaches, *J. of East China University of Science and Technology* (Natural Sci. Ed.), vol. 44 (5), pp. 744-751, 2018.
3. P. Chen and J Zhang, Performance comparison of machine learning algorithms for EEG-signal-based emotion recognition, *26th Int. Conf. on Artificial Neural Networks (ICANN17)*, 11-15 Sep. 2017, Alghero, Sardinia, Italy.

TOPIC 5: Intelligent Robotics

Pepper Humanoid Robots

- 1) Speech to text, 2) Text input to chatbot, 3) Chatbot response to speech
- Using different IBM Watson modules

Swarm Robots

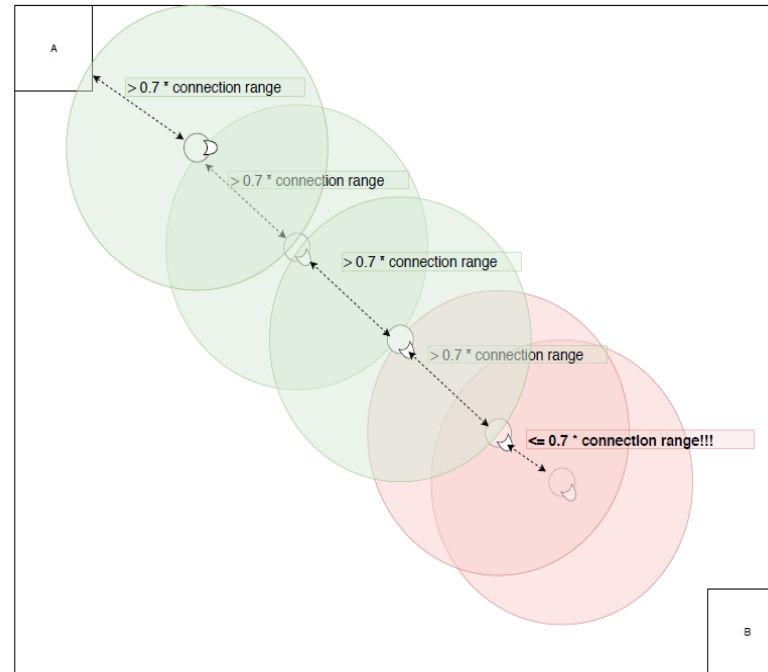
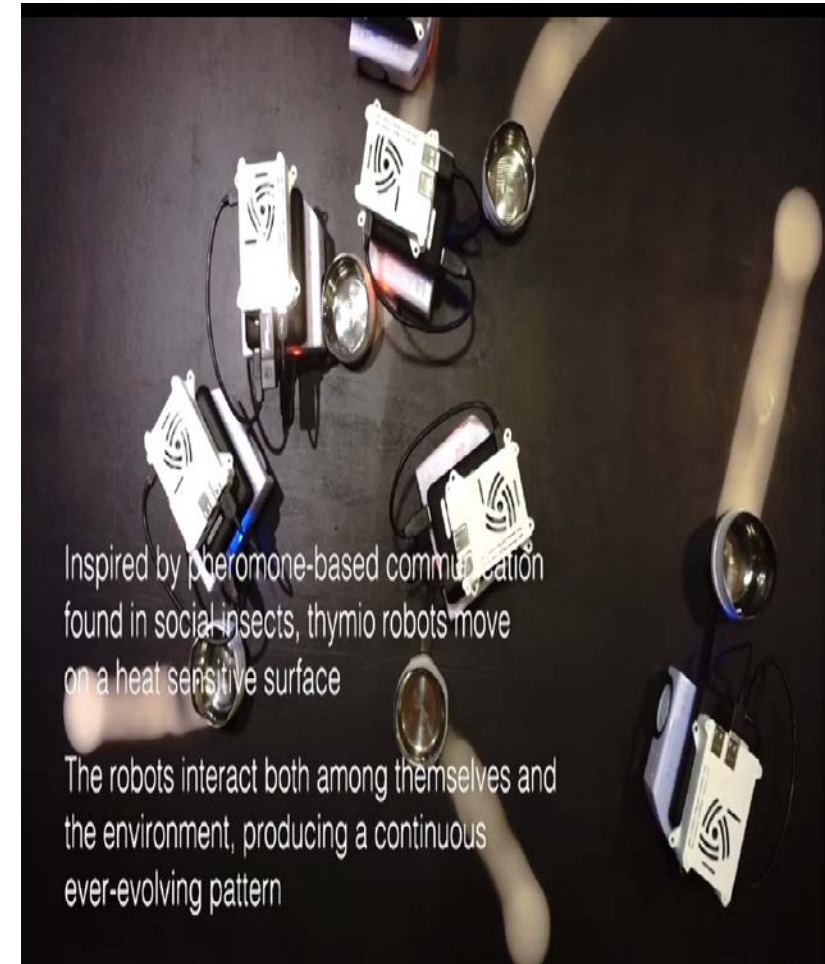
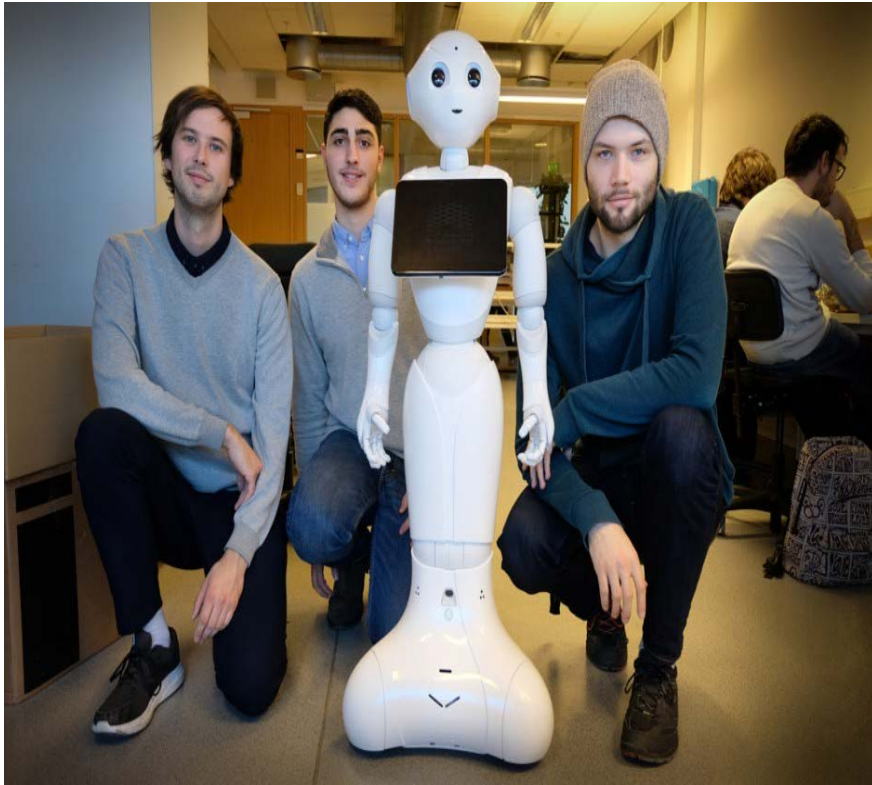


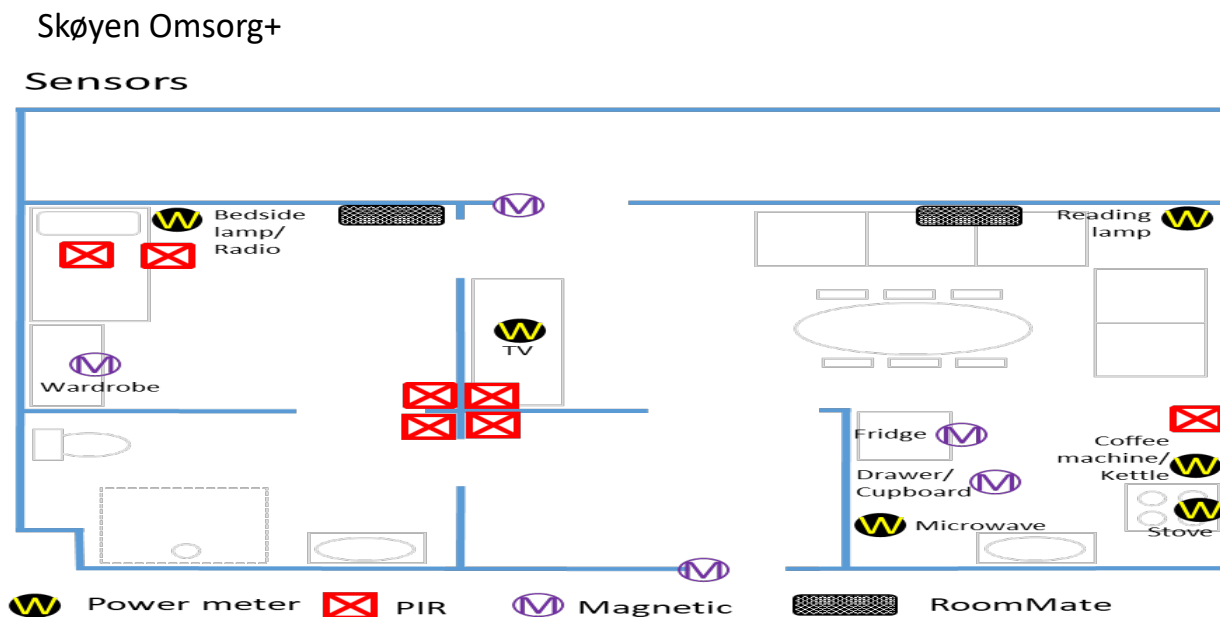
Fig. 6. Illustration of components involved in task being solved.



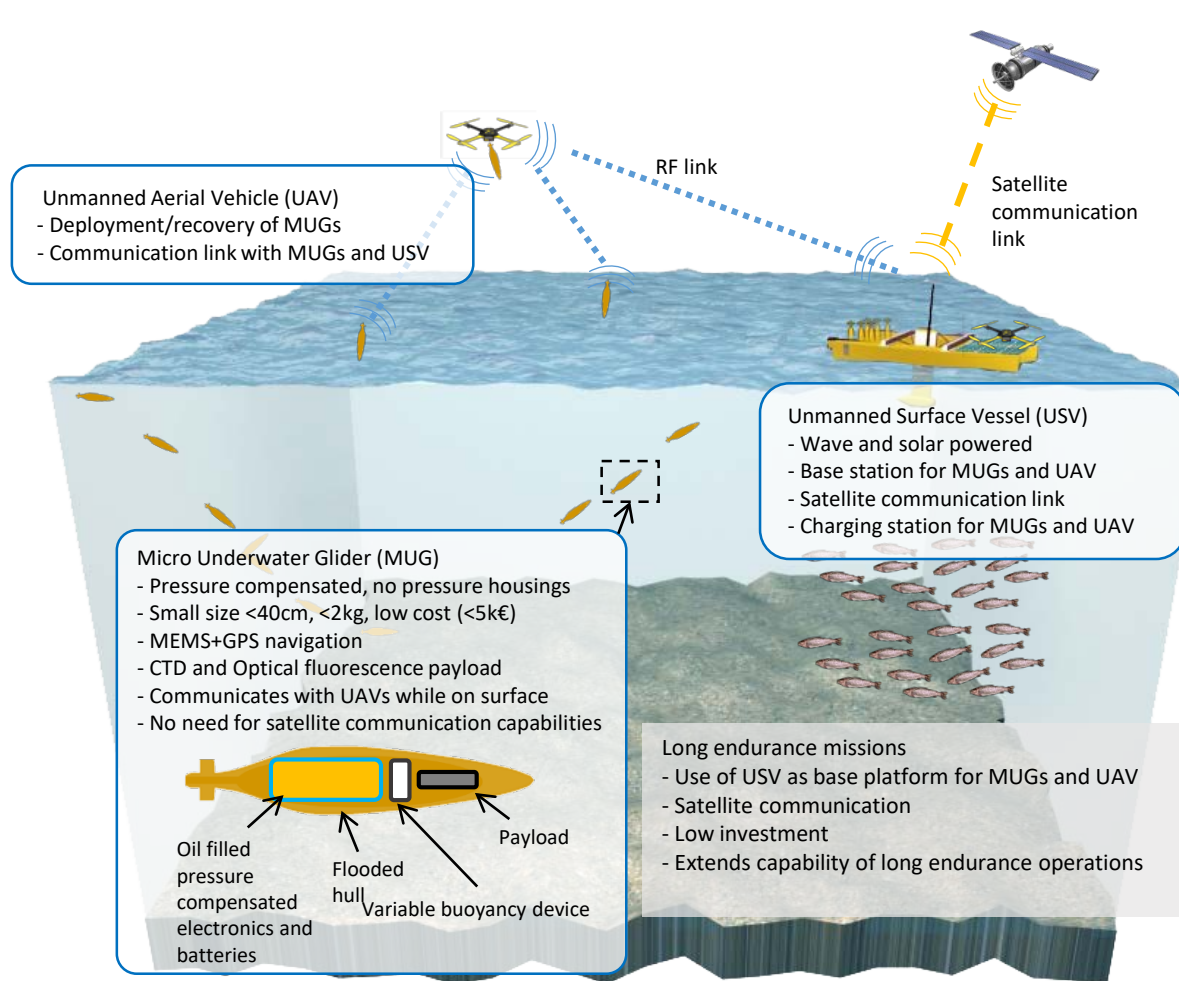
TOPIC 6: *Assisted Living Technology (ALT) for People with Mild Cognitive Impairment or **Dementia***



- Predict the next sensor to be activated/deactivated in a sequence.
- Useful for automation functions, such as:
 - turn on the coffee machine, when such event is predicted
 - turn on the lights in the kitchen, when the person wakes up at night



TOPIC 7: Ocean-Air synoptic operations using coordinated autonomous SYStems (OASYS)

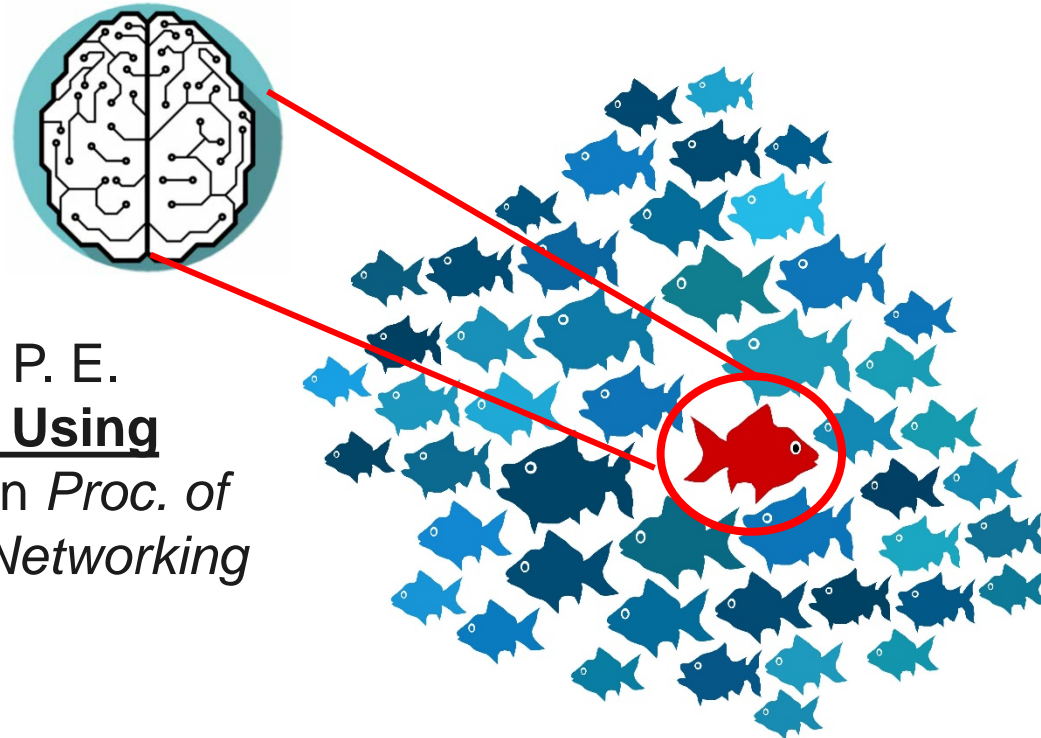


Funding: 1M €

Duration: 3 yrs

- Drones (UAVs): Deployment and recovery of gliders (MUG)
- Gliders (MUG) measure environmental parameters
- USV plays the role of mothership
- Drones (UAVs) are charged on board of the mothership

TOPIC 8: Intrusion Detection using ML



Hagos, D. H., Yazidi, A., Kure, Ø., & Engelstad, P. E. (2017), Enhancing Security Attacks Analysis Using Regularized Machine Learning Techniques. In *Proc. of 31st IEEE Int. Conf. on Advanced Information Networking and Applications (AINA)*, pp. 909-918.