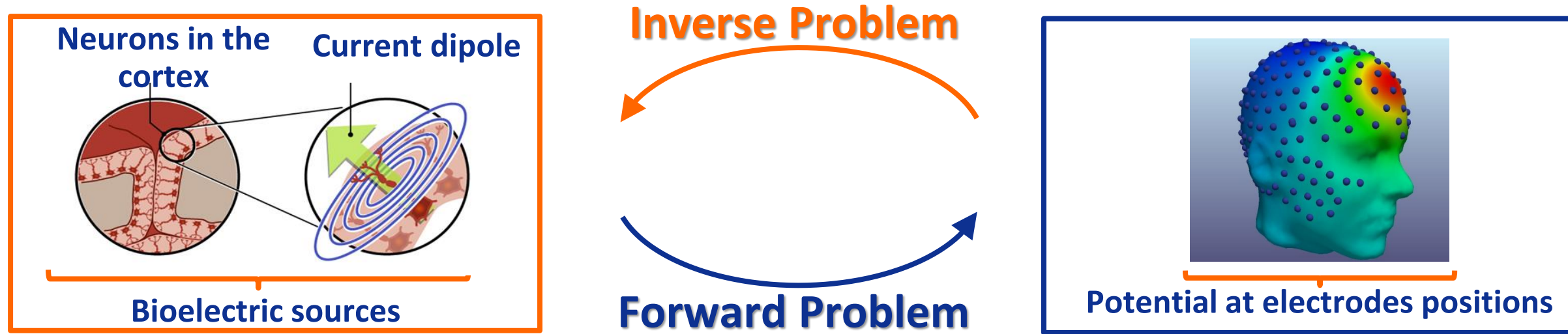


Research context and motivation

- **High resolution Electroencephalography (HR-EEG)** is a non-invasive, functional **neuroimaging technique**: it allows to determine position, strength and direction of the neural sources starting from the measure of the potential distribution at the scalp. HR-EEG is nowadays commonly used for the **pre-surgical epileptogenic source localization** in patients affected by focal epilepsy. Given the criticality of the applications, the source localization algorithms should provide **accurate** and **reliable** results in **short time**.
- EEG neuroimaging aims at solving an **inverse problem** (i.e. the identification of the neural sources from the measured scalp potential). It is commonly tackled by the iterative solution of the **forward EEG problem** (i.e. finding the scalp potential given a known source current inside the brain). Therefore, the accuracy and the resolution time of the forward problem affect the ones of the inverse problem. This work consists in the **stabilization** of a well-known integral formulation for the solution of the forward problem.



Addressed research questions/problems

- Under the **quasi-static** approximation, the problem is modeled by Poisson's equation, characterizing a **purely resistive model** of the head: $\nabla \cdot (\sigma \nabla V) = \nabla \cdot \mathbf{j}$.
- To recast this differential equation into a **boundary integral equation**, we model the biological tissues of the head, such as the brain, the skull and the scalp, as nested sets and assume **homogeneous conductivity** inside each. Moreover, we impose the continuity of the potential and of the current across the boundaries of these subdomains. We obtain the symmetric formulation:

$$Z \begin{pmatrix} \mathbf{v}^T \\ \mathbf{p}^T \end{pmatrix} = \begin{pmatrix} \mathbf{b}^T \\ \mathbf{c}^T \end{pmatrix}$$

$$Z = \begin{pmatrix} (\sigma_1 + \sigma_2)N_{11} & -\sigma_1 N_{12} & 0 & -2D_{11}^* & D_{12}^* \\ -\sigma_1 N_{21} & (\sigma_2 + \sigma_3)N_{22} & -\sigma_2 N_{23} & D_{21}^* & -2D_{22}^* \\ 0 & -\sigma_2 N_{32} & (\sigma_3)N_{33} & 0 & D_{32}^* \\ -2D_{11} & D_{12} & 0 & (\sigma_1^{-1} + \sigma_2^{-1})S_{11} & -S_{12}/\sigma_1 \\ D_{21} & -2D_{22} & D_{23} & -S_{21}/\sigma_1 & (\sigma_2^{-1})S_{22} \end{pmatrix}$$

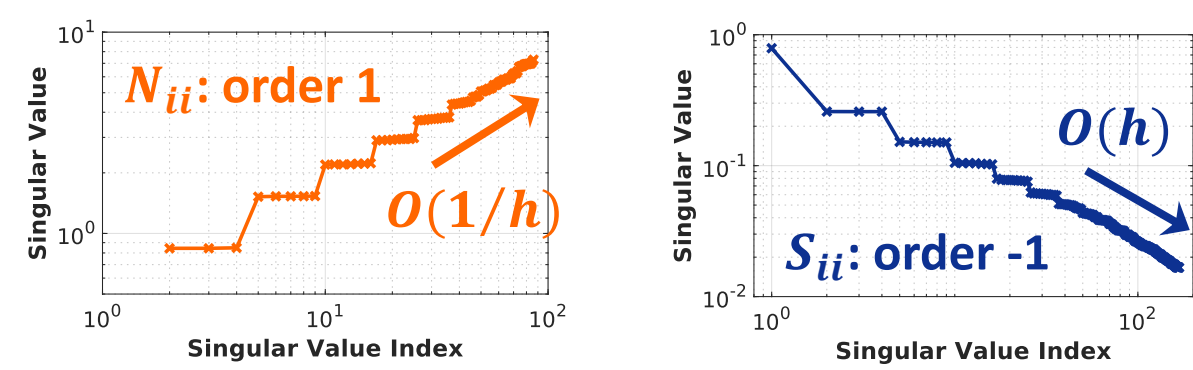
$$S_{ij}\psi(r) = \int_{\Gamma_j} G(r-r')\psi(r')dS(r')$$

$$D_{ij}\phi(r) = \text{p.v.} \int_{\Gamma_j} \partial_n G(r-r')\phi(r')dS(r')$$

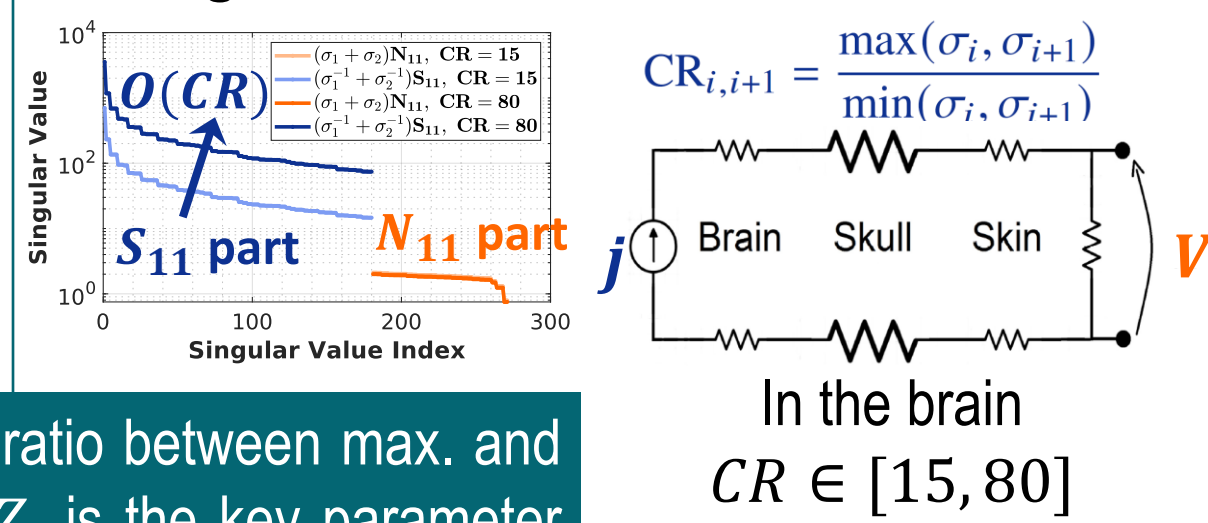
$$D_{ij}'\psi(r) = \text{p.v.} \int_{\Gamma_j} \partial_n G(r-r')\psi(r')dS(r')$$

$$N_{ij}\phi(r) = \text{f.p.} \int_{\Gamma_j} \partial_n \partial_n G(r-r')\phi(r')dS(r')$$

1. Dense-discretization breakdown



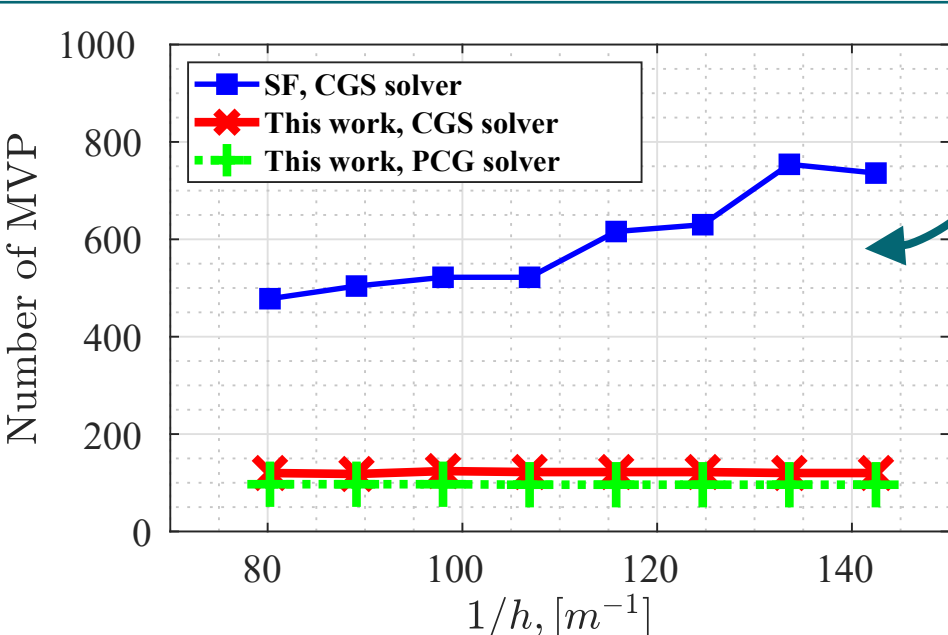
2. High-contrast breakdown



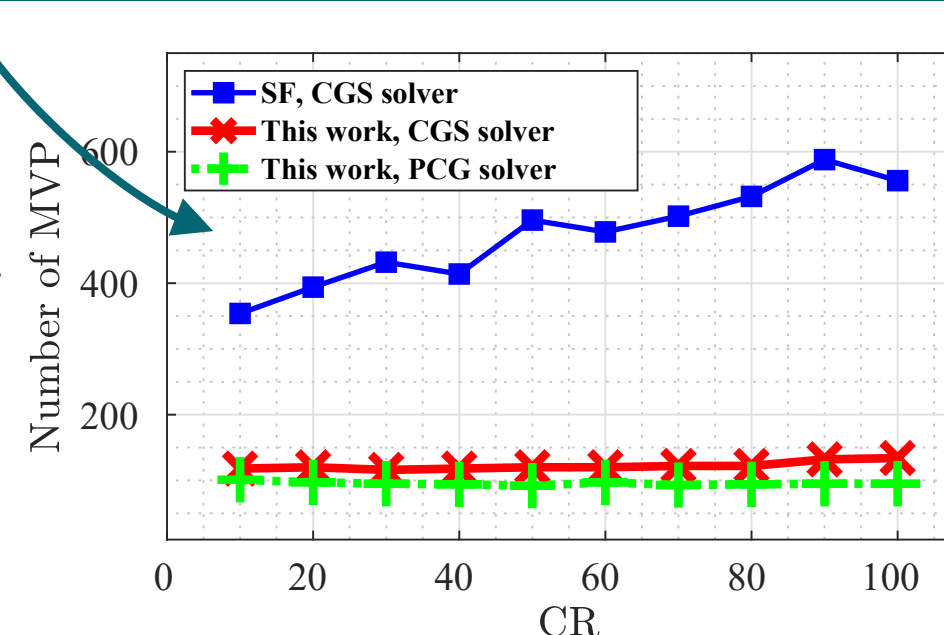
$cond(Z)$, given by the ratio between max. and min. singular values of Z , is the key parameter for evaluating the stability of the problem

$$cond(Z) = O(1/h^2)$$

$$cond(Z) = O(CR)$$



Increase of the number of iterations (or matrix-vector products) to solve the system iteratively.



Submitted and published works

- Giunzioni V., Ortiz G. John E., Merlini A., Adrian S.B., Andriulli F.P., "A New Refinement-Free Preconditioner for the Symmetric Formulation in Electroencephalography", IEEE AP-S, URSI 2022, Denver, 2022
- Giunzioni V., Ortiz G. John E., Merlini A., Adrian S.B., Andriulli F.P., "On a Refinement-Free Calderon Preconditioner for the Symmetric Formulation of the EEG forward problem", Journal of Computational Physics, submitted in Sept. 2022

Novel contributions

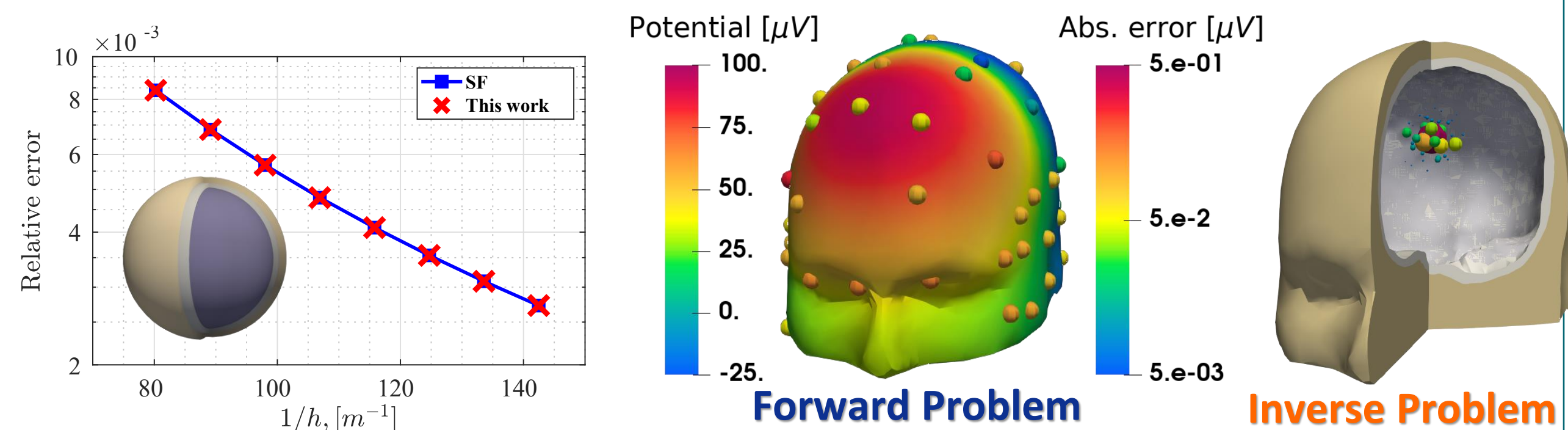
- Our new formulation gives rise to a **symmetric, positive-definite** interaction matrix.
 - Its construction is **refinement-free**: no need of dual functions defined on the barycentric refinement of the mesh.
 - The key idea: $N_{ii}\Delta_{\Gamma_i}^{-1}N_{ii}$ and $S_{ii}\Delta_{\Gamma_i}S_{ii}$ are **second-kind** integral operators, of pseudo-differential **order 0**, hence they are well-conditioned.
- $$Q = \text{diag}(q_{V,1}I_{N_{V,1}}, q_{V,2}I_{N_{V,2}}, \dots, q_{V,N}I_{N_{V,N}}, q_{C,1}I_{N_{C,1}}, q_{C,2}I_{N_{C,2}}, \dots, q_{C,N-1}I_{N_{C,N-1}})$$
- $$q_{V,i} = \max(\sigma_i, \sigma_{i+1})^{-1/2} \rightarrow Z_q = QZQ$$
- $$q_{C,i} = \min(\sigma_i, \sigma_{i+1})^{1/2}$$
- $$M = \text{diag}(G_{1,\pi\pi}^{-1/2}, G_{2,\pi\pi}^{-1/2}, \dots, G_{N,\pi\pi}^{-1/2}, G_{1,\lambda\lambda}^{-1/2}, G_{2,\lambda\lambda}^{-1/2}, \dots, G_{N-1,\lambda\lambda}^{-1/2})$$
- $$P = \text{diag}(\Delta_{1,\pi}^+, \Delta_{2,\pi}^+, \dots, \Delta_{N,\pi}^+, \tilde{\Delta}_{\pi,1}, \tilde{\Delta}_{\pi,2}, \dots, \tilde{\Delta}_{\pi,N-1})$$
- $$\tilde{\Delta}_{\pi,i} = G_{i,\lambda\pi}^{-1} \tilde{\Delta}_i G_{i,\pi\lambda}^{-1}$$
- $$(G_{i,jg})_{mn} = (f_{i,m}(r), \mathcal{I}g_{i,n}(r))_{L^2(\Gamma_i)}$$

Adopted methodologies

1. **Theoretical analysis**: the proposed preconditioning is based on **pseudo-differential operator theory**. Therefore, the **theoretical proof** of its good conditioning is based on the **spectral analysis** of each block, with the aim of determining its pseudo-differential order. Then, the analytic evaluation of the eigenvalues of the **principal part** of the formulation, independent of both mesh refinement and high contrast, ensures the **asymptotic boundness** of the condition number of the entire formulation, away from singularities.

$$Z_q P Z_q = \begin{pmatrix} \alpha(\sigma) \frac{N\Delta^+N}{\text{order } 0} + \beta(\sigma) \frac{D^*\tilde{\Delta}_\pi D}{\text{order } 0} & \gamma(\sigma) \frac{N\Delta^+D^*}{\text{order } -2} + \delta(\sigma) \frac{D^*\tilde{\Delta}_\pi S}{\text{order } 0} \\ \epsilon(\sigma) \frac{D\Delta^+N}{\text{order } -2} + \zeta(\sigma) \frac{S\tilde{\Delta}_\pi D}{\text{order } 0} & \eta(\sigma) \frac{D\Delta^+D^*}{\text{order } -4} + \theta(\sigma) \frac{S\tilde{\Delta}_\pi S}{\text{order } 0} \end{pmatrix}$$

2. **Numerical verification**: the stability of the formulation has also been shown **numerically**, on both canonical and realistic head models.



Future work

- **Acceleration** of the numerical solution of the preconditioned system by means of fast solvers, such as the **Adaptive Cross Approximation (ACA)**.
- Extension of the formulation to include realistic features of the brain models, such as **anisotropy** and **neuronal fibers self-interactions**.

List of attended classes

- 01DPJRV – Lens antennas: Fundamentals and present applications (7/12/2021, 10 h.)
- 01DOBRV – Mathematical-physical theory of electromagnetism (11/7/2022, 15 h)
- 01SFVRV – Metamaterials: Theory and multiphysics applications (1/4/2022, 20 h)
- 01UIZRV – Microwave sensing and imaging for innovative applications in health and food industry (22/3/2022, 20 h)
- 02LWHRV – Communication (24/2/2022, 5 h)
- 01SHMRV – Entrepreneurial Finance (25/2/2022, 5 h)
- 01UNVRV – Navigating the hiring process: CV, tests, interview (14/2/2022, 2 h)
- 01UNYRV – Personal branding (14/2/2022, 1 h)
- 08IXTRV – Project management (24/2/2022, 5 h)
- 01RISRV – Public speaking (10/2/2022, 5h)
- 01SYBRV – Research integrity (21/2/2022, 5 h)
- 01SWQRV- Responsible research and innovation (22/2/2022, 5 h)
- 02RHORV – The new Internet Society (24/2/2022, 6 h)
- 01UNXRV – Thinking out of the box (10/2/2022, 1h)
- 01SWPRV – Time management (22/2/2022, 2h)