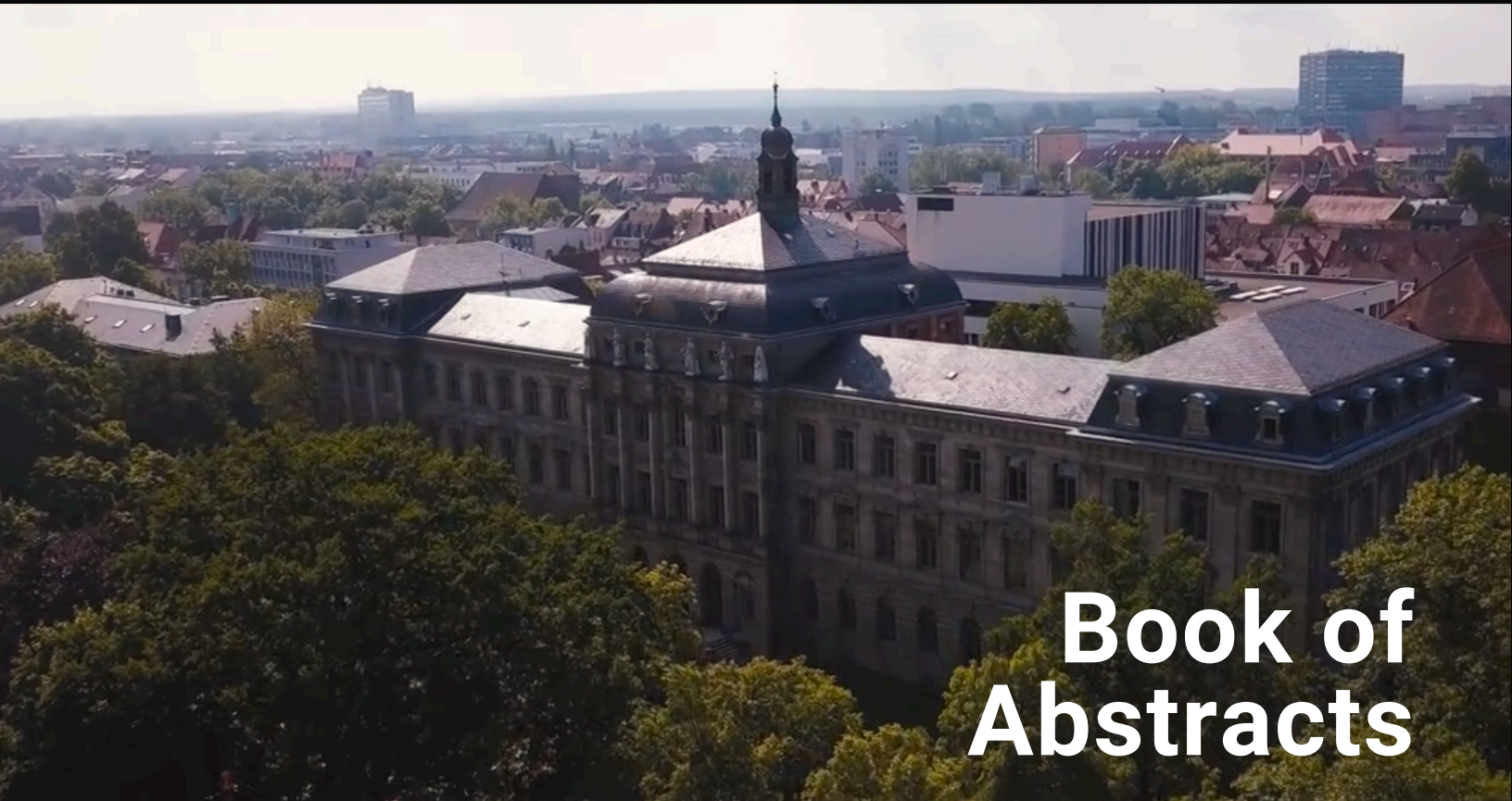


Machine Learning and PDEs

#MLPDES26 WORKSHOP

June 22 - 24, 2026

ERLANGEN, BAVARIA • ONLINE



Book of Abstracts



Friedrich-Alexander-Universität
Research Center for
Mathematics of Data | MoD

UNTERSTÜTZT VON / SUPPORTED BY



Alexander von
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Friedrich-Alexander-Universität
DYNAMICS, CONTROL,
MACHINE LEARNING
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MLPDES26

MACHINE LEARNING AND PDES
WORKSHOP 2026

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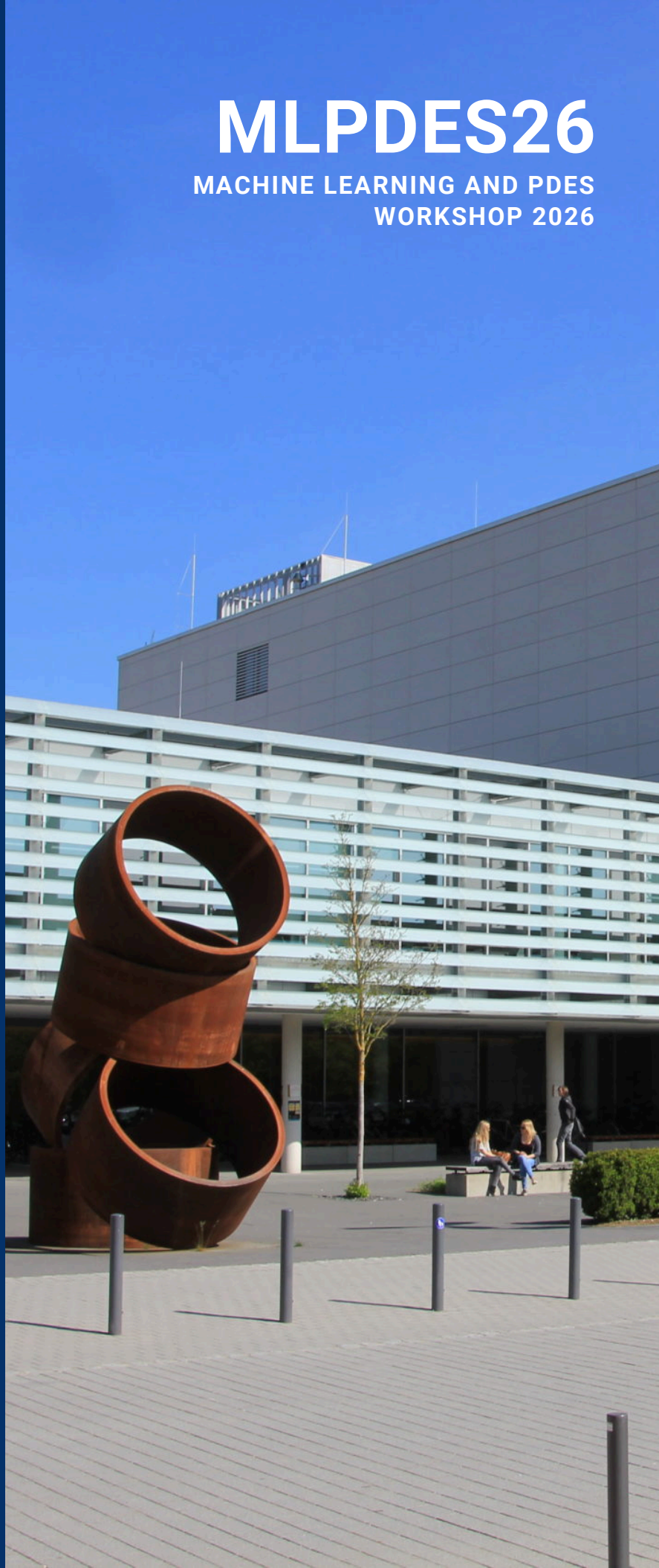
Scientific Committee's note

Invited Speakers

Program

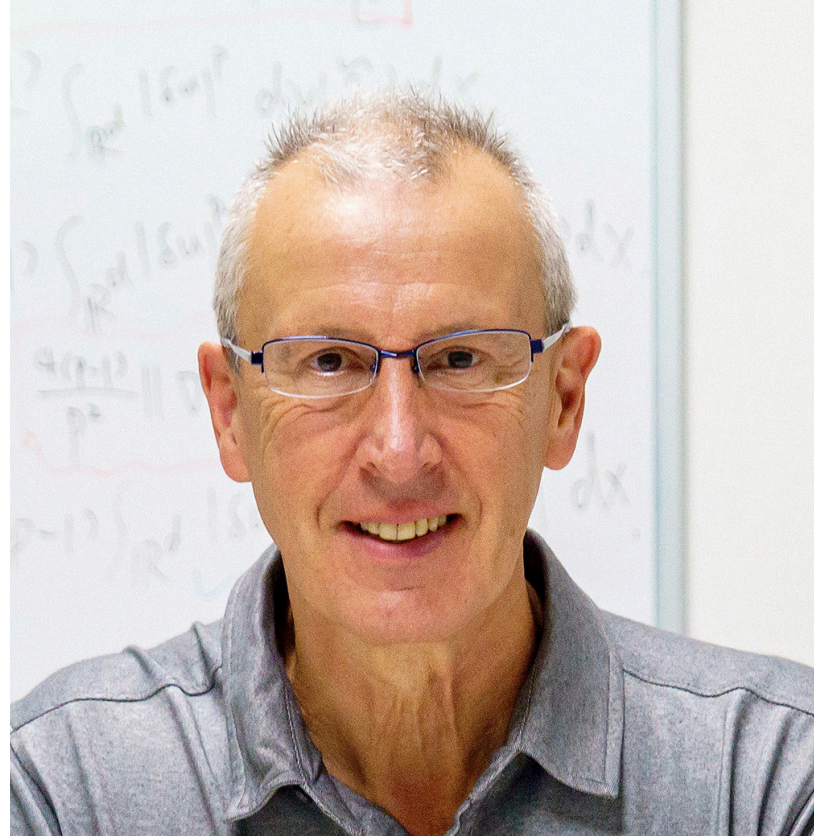
Abstracts

Organizing Committee



ERLANGEN • ONLINE | JUNE 22-24, 2026

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MLPDES26 Scientific Committee's Note

Hosted by FAU MoD at Friedrich-Alexander-Universität Erlangen-Nürnberg in Germany, the **#MLPDES26 workshop** is the second edition of an international event that brings together researchers from Europe, UK, China and the United States to explore the growing connection between Machine Learning (ML) and Partial Differential Equations (PDEs) –two core fields in modern mathematics that are now developing a dynamic, mutually beneficial relationship.

Supported by the FAU DCN-AvH, Chair for Dynamics, Control, Machine Learning and Numerics - Alexander von Humboldt Professorship, Politecnico di Bari and the Alexander von Humboldt Foundation, this event aims to establish a collaborative platform for participants from diverse

backgrounds to network, sharing insights, and driving progress in these exciting fields. Join us as we bridge these fields, focusing on both foundational research and practical applications.

See you in Erlangen & Online!

Giuseppe Maria Coclite

Politecnico di Bari

Enrique Zuazua

FAU MoD • FAU DCN-AvH
Alexander von Humboldt Professor
Friedrich-Alexander-Universität
Erlangen-Nürnberg



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MLPDES26

**MACHINE LEARNING AND PDES
WORKSHOP 2026**

Invited Speakers

ERLANGEN • ONLINE | JUNE 22-24, 2026

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Friedrich-Alexander-Universität
Research Center for
Mathematics of Data | MoD

Hosted by FAU MoD at Friedrich-Alexander-Universität Erlangen-Nürnberg, the **#MLPDES26 workshop** is the second edition of an international event that brings together researchers from Europe, UK and the United States to explore the growing connection between Machine Learning (ML) and Partial Differential Equations (PDEs) —two core fields in modern mathematics that are now developing a dynamic, mutually beneficial relationship.

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Machine Learning and PDEs

#MLPDES26 WORKSHOP

June 22 - 24, 2026

ERLANGEN - BAVARIA, GERMANY



WHEN

Mon.-Wed. **June 22 - 24, 2026**
09:30H - 17:00H

WHERE

On-site. FAU, Friedrich-Alexander-Universität Erlangen-Nürnberg
Senatssaal (Senate Hall) im Kollegienhaus
Universitätsstraße 15, 91054 Erlangen - Bavaria, Germany

Online (live streaming): fau.tv/clip/id/63196

REGISTRATION

Free but mandatory
Registration form: dcn.nat.fau.eu/registration-mlpdes26



mod.fau.eu/mlpdes26

SCIENTIFIC COMMITTEE

- **Giuseppe Maria Coclite.** Politecnico di Bari
- **Enrique Zuazua.** FAU, Friedrich-Alexander-Universität Erlangen-Nürnberg

ORGANIZING COMMITTEE

- **Darlis Bracho Tudares.** FAU, Friedrich-Alexander-Universität Erlangen-Nürnberg
- **Nicola De Nitti.** Università di Pisa
- **Lorenzo Liverani.** FAU, Friedrich-Alexander-Universität Erlangen-Nürnberg

AUDIENCE

This international workshop is open to: Public, Students, Postdocs, Professors, Faculty, Alumni and the scientific community all around the world



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June 22 - 24, 2026

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BAVARIA, GERMANY**

Machine Learning and PDEs

#MLPDES26 WORKSHOP

Invited Speakers



Riccardo Adami

PoliT0, Politecnico di Torino



Antonio Alvarez López

UAM, Autonomous University
of Madrid and FAU, Friedrich-
Alexander-Universität
Erlangen-Nürnberg



Harbir Antil

GMU, George Mason
University



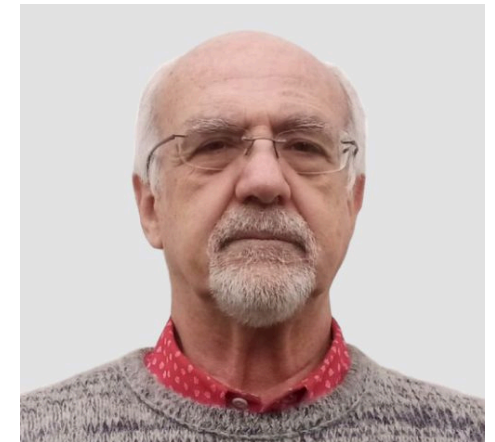
Luigi Berselli

Unipi, University of Pisa



Maria Bruna

University of Oxford



Giuseppe Buttazzo

Unipi, University of Pisa

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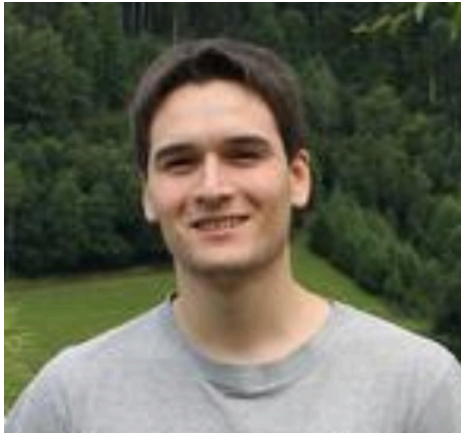
June 22 - 24, 2026

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Machine Learning and PDEs

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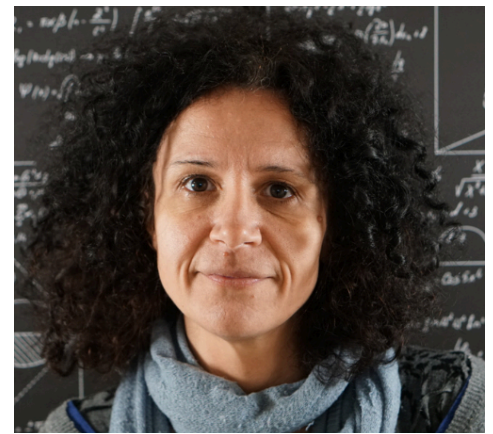
Gonzalo Cao Labora

EPFL, École Polytechnique
Fédérale de Lausanne



Elena Celledoni

NTNU, Norwegian University
of Science and Technology



Annalisa Cesaroni

Unipd, Università degli Studi di
Padova



Ulrik Skre Fjordholm

University of Oslo



Ingenuin Gasser

Universität Hamburg



Sam G. Krupa

ENS Paris, École Normale
Supérieure – PSL

ERLANGEN • ONLINE | JUNE 22-24, 2026

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Machine Learning and PDEs

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BAVARIA, GERMANY

Invited Speakers



Xue-Mei Li

EPFL, École Polytechnique
Fédérale de Lausanne



Francesco Maddalena

Poliba, Politecnico di Bari



Jean-Michel Morel

Lingnan University



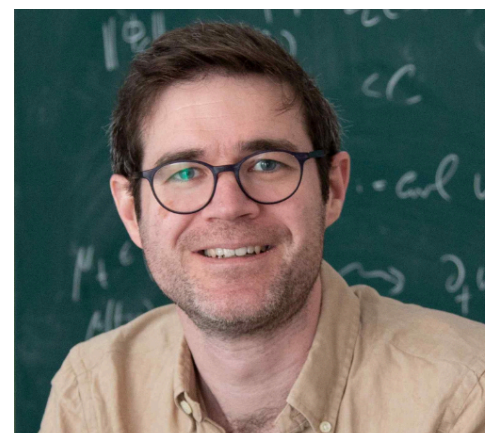
Lorenzo Pareschi

Heriot Watt University and
Università degli studi di
Ferrara



Yanzhi Wu

FAU DCN-AvH, Friedrich-
Alexander-Universität
Erlangen-Nürnberg



Emil Wiedemann

FAU, Friedrich-Alexander-
Universität Erlangen-Nürnberg

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Program

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Machine Learning and PDEs

#MLPDES26 WORKSHOP

June 22 - 24, 2026

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BAVARIA, GERMANY

#MLPDES26 Schedule • Time table

	MON. JUNE 22, 2026	TUE. JUNE 23, 2026	WED. JUNE 24, 2026
09:20H 09:50H	Registration Opening ceremony	Maria Bruna. Oxford Microscopic interactions and macroscopic PDEs: Beyond the mean-field paradigm	Emil Wiedemann. FAU Beyond Incompatibility: Using Optimal Transport for Algorithmic Fairness
10:00H 10:30H	Giuseppe Buttazzo. Unipi Optimal data for elliptic PDEs	Harbir Antil. GMU Digital Twins and Beyond: A PDE-Constrained Optimization Perspective	Sam G. Krupa. ENS-PSL Are L^∞ solutions to hyperbolic systems of conservation laws unique?
10:40H	COFFEE / TEA BREAK		
11:10H 11:40H	Elena Celledoni. NTNU Structure preservation and Deep Learning for Learning Mechanical Systems from Data	Ulrik Fjordholm. UIO The zero-noise limit for hyperbolic conservation laws	Yanzhi Wu. FAU DCN-AvH Data-Driven control of unknown linear systems: Two disturbance models
11:50H 12:20H	Xue-Mei Li. EPFL Rough PDEs, Long-Range Dependence, and Multi-Scale Dynamics	Riccardo Adami. PoliTO A mathematical model for the Einstein-Podolsky-Rosen phenomenon	Luigi Berselli. Unipi Womersley type flows for a non Newtonian fluid with variable exponent
12:30H	LUNCH BREAK		
14:00H 14:30H	Ingenuin Gasser. Uni-Hamburg Mathematical models in the context of renewable energies	Lorenzo Pareschi. HW Unife Structure-Preserving Neural Surrogates for Uncertainty Quantification in Plasma Physics	Closing ceremony
14:40H 15:10H	Gonzalo Cao-Labora. EPFL Discovery of unstable singularities	Francesco Maddalena. Poliba Multiple Scales in a Nonlocal Evolution Equation of Continuum Mechanics	
15:20H 16:00H	Jean-Michel Morel. Lingnan University. On Interpolation Formulas Describing Neural Network Generalization	Antonio Alvarez. UAM FAU Perceptrons and localization of attention's mean-field landscape	

SCIENTIFIC COMMITTEE

Giuseppe Maria Coclite. Poliba
Enrique Zuazua. FAU

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Darlis Bracho Tudares. FAU
Nicola De Nitti. Unipi
Lorenzo Liverani. FAU

WHERE

Onsite. FAU. Senate Hall. Kollegienhaus.
Universitätsstraße 15, 91054 Erlangen
Online (live-streaming): fau.tv/clip/id/63196



Machine Learning and PDEs

#MLPDES26 WORKSHOP

Program DAY 1 • Mon. June 22, 2026

09:20H

Registration

Opening ceremony

10:00H

10:30H



Giuseppe Buttazzo

Unipi, University of Pisa

Optimal data for elliptic PDEs

10:40H

COFFEE / TEA BREAK

11:10H

11:40H



Elena Celledoni

NTNU, Norwegian University of Science and Technology

Structure preservation and Deep Learning for Learning Mechanical Systems from Data

11:50H

12:20H



Xue-Mei Li

EPFL, École Polytechnique Fédérale de Lausanne

Rough PDEs, Long-Range Dependence, and Multi-Scale Dynamics

12:30H

LUNCH BREAK

14:00H

14:30H



Ingenuin Gasser

Universität Hamburg

Mathematical models in the context of renewable energies

14:40H

15:10H



Gonzalo Cao Labora

EPFL, École Polytechnique Fédérale de Lausanne

Discovery of unstable singularities

15:20H

16:00H



Jean-Michel Morel

Lingnan University

On Interpolation Formulas Describing Neural Network Generalization

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Lorenzo Liverani. FAU

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Program DAY 2 • Tue. June 23, 2026

09:20H
09:50H



Maria Bruna

University of Oxford

Microscopic interactions and macroscopic PDEs:
beyond the mean-field paradigm

10:00H
10:30H



Harbir Antil

GMU, George Mason University

Digital Twins and Beyond: A PDE-Constrained Optimization Perspective

10:40H

COFFEE / TEA BREAK

11:10H
11:40H



Ulrik Skre Fjordholm

University of Oslo

The zero-noise limit for hyperbolic conservation laws

11:50H
12:20H



Riccardo Adami

PoliTO, Politecnico di Torino

A mathematical model for the Einstein-Podolsky-Rosen phenomenon

12:30H

LUNCH BREAK

14:00H
14:30H



Lorenzo Pareschi

Heriot Watt University and Università degli studi di Ferrara

Structure-Preserving Neural Surrogates for Uncertainty
Quantification in Plasma Physics

14:40H
15:10H



Francesco Maddalena

Poliba, Politecnico di Bari

Multiple Scales in a Nonlocal Evolution Equation of
Continuum Mechanics

15:20H
16:00H



Antonio Alvarez

UAM, Autonomous University of Madrid and

FAU, Friedrich-Alexander-Universität Erlangen-Nürnberg

Perceptrons and localization of attention's mean-field landscape

Program DAY 3 • Wed. June 24, 2026

09:20H
09:50H



Emil Wiedemann

FAU, Friedrich-Alexander-Universität Erlangen-Nürnberg

Beyond Incompatibility: Using Optimal Transport for Algorithmic Fairness

10:00H
10:30H



Sam G. Krupa

ENS Paris, École Normale Supérieure – PSL

Are L^∞ solutions to hyperbolic systems of conservation laws unique?

10:40H

COFFEE / TEA BREAK

11:10H
11:40H



Yanzhi Wu

FAU DCN-AvH. Friedrich-Alexander-Universität Erlangen-Nürnberg

Data-Driven Control of Unknown Linear Systems: Two Disturbance Models

11:50H
12:20H



Luigi Berselli

Unipi, University of Pisa

Womersley type flows for a non Newtonian fluid with variable exponent

12:30H

LUNCH BREAK

14:00H
14:30H



Annalisa Cesaroni

Unipd, Università degli Studi di Padova

Analysis of commuting times in a linear city

14:40H

CLOSING CEREMONY

SCIENTIFIC COMMITTEE

Giuseppe Maria Coclite. Poliba
Enrique Zuazua. FAU

ORGANIZING COMMITTEE

Darlis Bracho Tudares. FAU
Nicola De Nitti. UniPi
Lorenzo Liverani. FAU

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MACHINE LEARNING AND PDES
WORKSHOP 2026



Abstracts

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Abstracts

**Giuseppe Buttazzo** Unipi, University of Pisa
Optimal data for elliptic PDEs

The goal is to present some optimization problems related to an elliptic PDE, where the control variables are the data of the equation. The cases we consider are the following.

- The PDE is of the form $-\operatorname{div}(a(x)\nabla u)=f$ and the control is the coefficient a ;
- The PDE is of the form $-\Delta u+V(x)u=f$ and the control is the potential V ;
- The PDE is of the form $-\Delta u=f$ and the control is the right-hand side f .

In all the cases the boundary conditions are of Dirichlet type, and the cost functional is an integral like $J(u)=\int_{\Omega} j(x,u,\nabla u)\,dx$. The admissible controls are constrained to satisfy an integral condition as $\int_{\Omega}\psi(a)\,dx\leq m$, $\int_{\Omega}\psi(V)\,dx\leq m$, $\int_{\Omega}\psi(f)\,dx\leq m$ respectively. The goal is to obtain, under suitable assumptions on the functions j and ψ , the existence of optimal controls. The results are in a series of works in collaboration with J. Casado-Díaz and F. Maestre, from Universidad de Sevilla.

**Elena Celledoni** NTNU, Norwegian University of Science and Technology
Structure preservation and Deep Learning for Learning Mechanical Systems from Data

In this talk I will review work on the analysis of motion capturing data and similar applications using techniques of shape analysis and deep learning. I will then consider a method for learning the Lagrangian and forces for mechanical systems using the discrete Lagrange d'Alembert principle. The case of manifold valued data and data on Lie groups will also be discussed if time permits. Applications to mechanical system will be considered.

**Xue-Mei Li** EPFL, École Polytechnique Fédérale de Lausanne
Rough PDEs, Long-Range Dependence, and Multi-Scale Dynamics

I will discuss the role of long-range dependence and multiple time scales in partial differential equations. In particular, I will focus on stochastic dynamics driven by long-range dependent noise. Whether machine learning can contribute to the understanding of such systems, especially in the presence of multiple time scales, remains an open and intriguing question.

Abstracts



Ingenuin Gasser Universität Hamburg
Mathematical models in the context of renewable energies

The presentation refers to the mathematical modelling of power plants based on renewable resources, ranging from established applications such as parabolic trough power plants to osmotic pressure driven energy conversion technologies. The complexity of such applications requires a substantial mathematical modelling effort to finally end up with reasonable models, which can be simulated fast and robust and which allow optimisation approaches. All applications involve fluid dynamic or thermo-fluid dynamic models which have to be significantly reduced under the restriction of keeping the most relevant chemical, bio-chemical and physical effects. Mathematically these models mostly are nonlinear systems of ODEs or PDEs of mixed elliptic-hyperbolic type. We present the most recent results on some of the models. In view of the underlying application we discuss which quantities are reasonable to be optimized, e.g. the power output of the power plant with respect to operational or system parameters.



Gonzalo Cao Labora EPFL, École Polytechnique Fédérale de Lausanne
Discovery of unstable singularities

I will present a new method based on physics-informed neural networks that allows us to discover singularities in fluid PDEs. Our setting focuses on self-similar solutions—those that evolve in time solely through horizontal and vertical rescalings. In this framework, the problem of finding singularities reduces to a PDE without initial data nor time dependence, and the central challenge becomes determining if any smooth solutions to this equation exist. We will see how neural networks are able to efficiently explore the space of functions to find these solutions. This is particularly valuable in the case of unstable solutions, where traditional numerical methods often struggle. Specifically, I will present the discovery of new solutions for the incompressible porous media equation, the Euler equations with boundary, and the Cordoba-Cordoba-Fontelos model. Time permitting, we will also discuss the use of computer-assisted techniques to rigorously prove that our numerically found solutions are close to exact, true solutions. Joint work with Yongji Wang, Tristan Buckmaster, Javier Gomez-Serrano, Ching-Yao Lai, and the team from Google DeepMind.

Machine Learning and PDEs

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Jean-Michel Morel Lingnan University

On Interpolation Formulas Describing Neural Network Generalization

In 2020, Pedro Domingos proposed an interpolation formula that is valid for any model learnt using deterministic gradient descent. This formula expresses the output of a trained machine at a test point as a weighted sum of contributions from all the training samples along the optimisation path. These weights are the scalar products of the machine's gradients with respect to its parameters, computed at the test and training data respectively. This result, which was proven for deterministic gradient descent, supports the idea that learning involves storing data-dependent information in an evolving tangent feature geometry. Predictions at test time then arise from the retrieval and aggregation of these stored features using a kernel weighting. Generalisation is governed by the alignment between the test points and the learned feature memory. We will demonstrate that Domingos's interpolation formula can be extended to batch-based stochastic optimisation. First-order interpolation formulas are obtained in terms of the learning rate, with a form that varies depending on the type of stochastic optimisation performed: gradient descent (GD), stochastic gradient descent (SGD), and stochastic gradient descent with momentum (SGDM). These formulas have a second-order counterpart that captures how local loss curvature and mini-batch fluctuations further refine the prediction under finite learning rates. We validate the theory through experiments, demonstrating that second-order representations provide a more accurate description of test output than the first-order approximation.



Maria Bruna University of Oxford

Microscopic interactions and macroscopic PDEs: beyond the mean-field paradigm

Many PDE models arising in collective behaviour, optimisation, and machine learning can be interpreted as limits of interacting particle systems. While mean-field limits are by now well understood and widely used, they capture only one regime of interaction. In this talk, I will present a broader perspective on particle-to-continuum limits, focusing on how different interaction scales lead to different classes of PDEs. Weak, long-range interactions give rise to nonlocal mean-field equations, whereas short-range or localised interactions lead to nonlinear diffusion-type equations. In contrast, strongly interacting systems, such as those with exclusion or finite-size effects, can exhibit macroscopic behaviour that deviates significantly from mean-field predictions, particularly in multi-species or non-gradient settings. Using examples from both lattice models and interacting diffusions, I will highlight the mechanisms behind these limits and discuss what carries over (and what breaks down) between regimes. This viewpoint suggests new challenges for connecting microscopic models with PDEs used in machine learning and related areas.

Machine Learning and PDEs

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Harbir Antil GMU, George Mason University

Digital Twins and Beyond: A PDE-Constrained Optimization Perspective

Digital Twins (DTs) are adaptive, real-time virtual replicas of physical systems that integrate physics-based models, sensor data, and intelligent decision-making. At their core, DTs can be rigorously framed within PDE-constrained optimization (PDECO). This talk develops a unified PDECO framework for state estimation and control, leveraging adjoint-based methods in both deterministic and stochastic settings. To address the challenges of infinite-dimensional, large-scale optimization, we introduce novel function-space trust-region and augmented Lagrangian algorithms, and explore the role of randomized methods in dynamic PDECO. Applications span a wide range of domains, including structural and biomedical systems—from bridges and dams to aneurysm modeling, optimal insulation, electromagnetic cloaking, light bending, fusion, and neuromorphic computing. Together, these examples highlight a pathway toward predictive, adaptive, and trustworthy Digital Twins and AI technologies.



Ulrik Skre Fjordholm University of Oslo

The zero-noise limit for hyperbolic conservation laws

Over the past two decades, the phenomenon of regularization by noise has received an enormous amount of attention. Put simply, this is a general phenomenon in which a stochastic differential equation might exhibit better properties — e.g., existence or uniqueness of solutions, or increased regularity — compared to its deterministic counterpart. An early example is the seminal work of Flandoli, Gubinelli & Priola (*Inventiones Mathematicae*, 2010), where the authors prove that a specific stochastically perturbed transport equation is well-posed even for very irregular (Hölder-continuous) velocity fields. The purpose of this work is threefold. First, we design a novel stochastic perturbation of general scalar conservation laws, and prove that this SPDE is well-posed, without imposing any entropy conditions. Second, we show that there is sufficient compactness to pass the noise parameter to zero. Third, we use a novel stability property of scalar conservation laws to prove that the zero-noise limit is in fact the entropy solution of the deterministic problem. In this way, the addition of (a vanishing amount of) noise acts as a selection mechanism, which coincides with the standard entropy conditions. This is joint work with Ola Mæhlen (Université Paris-Saclay) and Magnus Ch. Ørke (University of Oslo).

Abstracts



Riccardo Adami PoliT0, Politecnico di Torino

A mathematical model for the Einstein-Podolsky-Rosen phenomenon

At the core of the second technological quantum revolution is a physical phenomenon known as "entanglement". First figured out by Schrödinger, it was employed in a famous article by Einstein, Podolsky and Rosen to show the incompleteness of Quantum Mechanics. The interest on entanglement arose then again in the sixties, and since then it has been deeply investigated, both theoretically and experimentally. In this talk, we construct a mathematical model that describes the evolution of two entangled particles and a spin, that interacts with one of the two. As a result, we give a quantitative description of the phenomenon of the entanglement and, owing to a suitable scaling, we single out the regime in which the phenomenon occurs. This is a joint work with Luigi Barletti and Alessandro Teta.



Lorenzo Pareschi Heriot Watt University and Università degli studi di Ferrara

Structure-Preserving Neural Surrogates for Uncertainty Quantification in Plasma Physics

Reliable uncertainty quantification is a central challenge in plasma simulation, especially in kinetic regimes where predictive computations remain extremely expensive. This issue is particularly relevant in fusion-oriented plasma modeling, where multiscale effects, high dimensionality, and sensitivity to uncertain inputs make brute-force sampling unaffordable. In this talk I will present a multifidelity framework for the Vlasov-Poisson-Landau equation that combines asymptotic-preserving solvers, reduced plasma models, and tensor neural surrogates based on a micro-macro decomposition. The resulting approach produces inexpensive low-fidelity samples that remain strongly correlated with the high-fidelity kinetic model, leading to substantial variance reduction and computational savings. More broadly, this provides an example of how machine learning can become genuinely effective for PDEs when it is built around structure rather than used as a black box.



Francesco Maddalena Poliba, Politecnico di Bari

Multiple Scales in a Nonlocal Evolution Equation of Continuum Mechanics

The analysis of solutions of a linear peridynamic equation reveals some interesting features due to the particular non-homogeneous dispersion relation, which constitutes the main source of a nontrivial scale-dependent behaviour. The talk will focus on a quantitative analysis of these aspects, also in the perspective of their physical relevance.

Machine Learning and PDEs

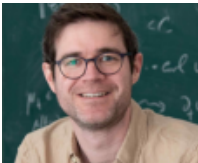
#MLPDES26 WORKSHOP

Abstracts



Antonio Alvarez UAM, Autonomous University of Madrid and FAU,
Friedrich-Alexander-Universität Erlangen-Nürnberg
Perceptrons and localization of attention's mean-field landscape

Self-attention, the core mechanism behind modern Transformers, admits a natural mean-field description as a gradient flow on the sphere. In this talk, I will discuss what happens when attention is coupled with a perceptron term, leading to a simplified mean-field model for Transformer layers. The main message is that the perceptron induces a localization mechanism: under suitable structural assumptions, stationary measures are forced to be singular, and in dimension two they are in fact purely atomic with finite support. Thus, even in the descent regime, where attention alone favors diffuse equilibria, the perceptron can enforce clustered stationary states. I will also discuss anti-concentration estimates in the descent regime, showing that these discrete stable equilibria cannot concentrate too much mass inside a small cluster.



Emil Wiedemann FAU, Friedrich-Alexander-Universität Erlangen-Nürnberg
Beyond Incompatibility: Using Optimal Transport for Algorithmic Fairness

Automated decision-making has been proposed and used, for instance, in college admissions, credit scoring, and even criminal justice. It was originally hoped to yield fairer results by avoiding human biases, and thus to minimise effects of discrimination. Of course this turned out wrong ("bias in, bias out"), sparking serious concern in the public and media. Even worse, it turns out that different, equally plausible fairness criteria can be fundamentally incompatible with each other. Indeed, a strongly politicised debate about racial discrimination through the COMPAS algorithm in US criminal justice essentially boils down to such an incompatibility, as shown in a highly cited paper by Kleinberg et al. (2016). I will give a complete and very elementary proof of their incompatibility theorem, and discuss how such incompatibilities can be mitigated via optimal transport methods. Joint work with Meike Zehlike and Philipp Hacker.



Sam G. Krupa ENS Paris, École Normale Supérieure – PSL
Are L^∞ solutions to hyperbolic systems of conservation laws unique?

For hyperbolic systems of conservation laws in 1-D, fundamental questions about uniqueness and blow up of weak solutions still remain even for the apparently "simple" systems of two conserved quantities such as isentropic Euler and the p-system. Similarly, in the multi-dimensional case, a longstanding open question has been the uniqueness of weak solutions with initial data corresponding to the compressible vortex sheet. We address all of these questions by using the lens of convex integration, a general method of constructing highly irregular and non-unique solutions to PDEs. Our proofs involve computer-assistance. This talk is based on joint work with László Székelyhidi, Jr.

Machine Learning and PDEs

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Abstracts



Yanzhi Wu FAU DCN-AvH Friedrich-Alexander-Universität Erlangen-Nürnberg
Data-Driven Control of Unknown Linear Systems: Two Disturbance Models

This talk discusses data-driven control of unknown linear systems under two disturbance models. For exosystem-generated disturbances, the framework is optimal output regulation: regulator equations characterize the required steady-state response, and adaptive dynamic programming learns the feedback component while feedforward regulation is obtained from the regulator equations. For stochastic disturbances with distributional ambiguity, the framework becomes mean-penalized Wasserstein distributionally robust control: a finite-horizon minimax Bellman recursion is learned from input-output data using a history-based state representation and quadratic Q-function parameterization. The talk highlights how the disturbance model determines the mathematical object, the learning algorithm, and the resulting regulation or robust-performance guarantee.



Luigi Berselli Unipi, University of Pisa
Womersley type flows for a non Newtonian fluid with variable exponent

We study the fully-developed, time-periodic motion of a shear-dependent non-Newtonian fluid with variable exponent rheology through an infinite pipe, with Dirichlet conditions. The main new results concern $S_p(\cdot)$ -fluid models. We identify classes of exact solutions, relevant as benchmark cases, especially for electro-rheological fluids together with abstract existence results confirmed by numerical experiments.



Annalisa Cesaroni Unipd, Università degli Studi di Padova
Analysis of commuting times in a linear city

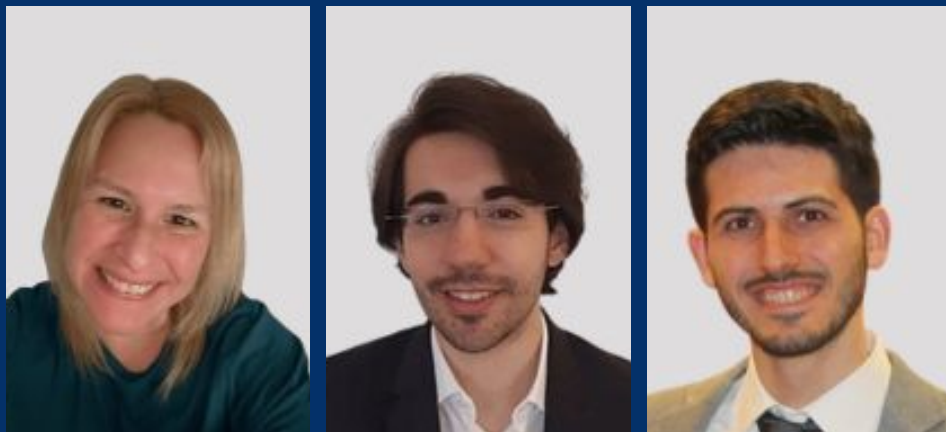
In this talk, I will provide a mathematical model (in a simplified framework) that describes the daily phenomenon in which each of thousand of citizens chooses to commute from a given location to another one, starting at a given time and somehow guessing the arrival time. In the discrete setting, I will use the celebrated Follow-the-Leader model, with the original feature of cars entering and exiting at different times, and I will provide the convergence as the number of players increase to infinity to a balance law in the continuous setting. Based on joint works (in progress) with F. Ancona, L. Caravenna, E. Marconi, A. Marson from University of Padova and with F. Rossi from IUAV Venice.



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