Dr. rer. nat. Sebastian Lapuschkin

* December 16, 1986 in Würzburg

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 $github.com/sebastian-lapuschkin \bullet linkedin.com/in/sebastian-lapuschkin scholar.google.com/citations?user=wpLQuroAAAAJ$

Short Bio

Sebastian Lapuschkin is the Head of the Explainable Artificial Intelligence research group at Fraunhofer Heinrich Hertz Institute (HHI) in Berlin.

He received his Ph.D. degree with distinction from the Technische Universität Berlin in 2018 for his pioneering contributions to the field of Explainable Artificial Intelligence (XAI) and interpretable machine learning. From 2007 to 2013 he studied computer science (B. Sc. and M. Sc.) at the Technische Universität Berlin, with a focus on software engineering and machine learning.

Sebastian is the recipient of multiple awards, including the Hugo-Geiger-Prize for outstanding doctoral achievement and the 2020 Pattern Recognition Best Paper Award.

His research has shaped the field of XAI from the very beginning, with contributions to the first wave of XAI such as the popular and widely-used Layer-wise

Relevance Propagation method, as well as timely works influencing the second wave of XAI with additions to the sub-fields of Mechanistic Interpretability, Data Attribution and XAI-based model- and data improvement.

Sebastian is an avid advocate for Open Science, demonstrated by numerous Free Open Source Software toolboxes published with the intent to warrant and facilitate reproducibility in AI research.

Since 2024 he is co-organizing The World Conference on eXplainable Artificial Intelligence and serves as a Topic Editor on "Opportunities and Challenges in Explainable Artificial Intelligence" for the MDPI Open Access Journals.

Further research interests include efficient machine learning and data analysis, as well as data and algorithm visualization.

Professional Experience

Technological University Dublin

Dublin, Ireland 2025 -

External Scholar

at the Centre of eXplainable Artificial Intelligence. The Centre is the first of its kind in the Republic of Ireland and it aims to increase further and synergise cross-centres, college and external research collaboration.

Multidisciplinary Digital Publishing Institute (MDPI)

Topic Editor 2024 - 2026

for "Opportunities and Challenges in Explainable Artificial Intelligence". Submission pre-screening, review management and decision handling.

XAI4Science

Organizer 2024 -

of the 1st workshop "XAI4Science: From Understanding Model Behavior to Discovering New Scientific Knowledge (2025)", co-located with ICLR 2025 at Singapore EXPO, Singapore. TBA: 2nd of the workshop at AAAI 2026, Singapore.

World Conference on eXplainable Artificial Intelligence

Steering Committee Member

2024 -

for the 3rd XAI World Conference (2025) in Istanbul, Turkey, as well as future undisclosed instances of the conference. Panelist, Conference and Special Track co-organization.

Programme Committee Chair

2023 - 2024

for the 2nd XAI World Conference (2024) in Valetta, Malta.

Conference and Special Track co-organization.

Fraunhofer Heinrich-Hertz-Institute

Berlin, Germany

2025 -

for the Erasmus Mundus Joint Master in Intelligent Field Robotic Systems (IFRoS) associate partnership of Fraunhofer HHI.

Contact Person

Ethics Committee Member Founding member of the first ethics committee at Fraunhofer HHI.	2023 -
Head of Explainable Artificial Intelligence Research Group Leadership and direction of XAI research & applications.	2021 -
Tenured Researcher PostDoc research position in the Machine Learning Group at Fraunhofer HHI.	2019 - 2020
Research Associate Founding member of the Machine Learning Group at Fraunhofer HHI.	2014 - 2018
Technische Universität Berlin Research Associate Supervision by Prof. Dr. Klaus-Robert Müller.	Berlin, Germany 2013 - 2014
Student Research- & Teaching Assistant Research & Teaching assistant to Prof. Dr. Klaus-Robert Müller.	2011 - 2013
Student Teaching Assistant Teaching assistant to Prof. Dr. Marc Alexa, Prof. Dr. Odej Kao and Prof. Dr. Oliver Brock	2009 - 2011 s.
Education	
Technische Universität Berlin PhD in Machine Learning (summa cum laude) Date of oral defense: December 19 th , 2018. Dean's signature on Doctorate Certificate dated January 23 rd , 2019.	Berlin, Germany 2018
Thesis: "Opening the machine learning black box with Layer-wise Relevance Propagation Supervision headed by Prof. Dr. Klaus-Robert Müller.	ı"
Master of Science in Computer Science Focus on machine learning, computer vision and large scale data analysis.	2013
Bachelor of Science in Computer Science Focus on algorithms and software development	2010
Deutschhaus-Gymnasium Abitur (pre-university secondary education) With the secondary education with th	ürzburg, Germany 2007
Teaching Teaching of and teaching support for 19 university courses since 2009, including	
"Responsible Artificial Intelligence 1", Technische Universität Berlin	2023 -
"(An Introduction to) Explainable AI", Universitat de Girona / IFRoS	2021 -
Talks & Lectures	
Over 30 invited talks and individual lectures held since 2017, including Explainability in the Era of LLMs: New Challenges and Pathways to Actionable Insigl 2025 Workshop on Self-Supervised Learning for Signal Decoding, Aalborg, Denmark	hts 2025
XAI as a Tool Beyond Model Understanding – From Heatmaps to Concepts and XAI A CBS CoCoNUT Max Planck Institute for Human Cognitive and Brain Sciences	utomation 2024
Explaining AI with Concept Relevance Propagation $4^{\rm th}$ Japanese-American-German Frontiers of Science (JAGFOS) Symposium	2023
Towards Human-understandable Explanations with XAI 2.0 AI4Good webinar series of the International Telecommunication Union (ITU)	2022
Towards Actionable XAI International Artificial Intelligence Doctoral Academy	2022

Third-Party Funded Research Projects

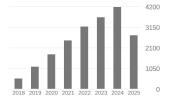
7 third-party funded research projects acquired and managed since 2018, including

ACHILLES Human-Centred Machine Learning: Lighter, Clearer, Safer Funded with 8.2MM€ by the European Union	2024 - 2028
TEMA Trusted Extremely Precise Mapping and Prediction for Emergency Management Funded with 11.3MM€ by the European Union	2022 - 2026
iToBoS	2021 - 2025
Intelligent Total Body Scanner Funded with 11.7MM€ by the European Union	
Honors & Awards	
Machine Learning and the Physical Sciences Reproducibility Badge	2024
Stanford Top 2% Scientist Worldwide*	2021 - 2023
Best Short Paper Award	2023
Pattern Recognition Best Paper Award and Pattern Recognition Medal	2020
Hugo-Geiger-Prize (1st place)	2019
Freunde des HHI Nachwuchspreis	2019
ERCIM Cor van Baayen Award (finalist)	2019
Best Paper Award	2016
Patents	
Attention Head Functionalities in Machine Learning Models	2025
Relevance Score Assignment dealing with an Attention Module and Applications thereof	2024
A Concept Representation of a Machine Learning Model	2024
Analyzing an Inference of a Machine Learning Predictor	2023
Method and System for Simulating an Optical Image of a Photonic and/or Electronic Device	2022
Pruning and/or Quantizing Machine Learning Predictors	2020
Relevance Score Assignment for Artificial Neural Networks	2016

Publications

Summary of Scientific Impact

	All	Since 2020
# Publications	95	75
# Citations	20230	18051
h-index	37	36
i10-index	62	61



per Google Scholar, retreived on September 1^{st} , 2025.

Selected Publications

Theoretical & Methodological Contributions to XAI

1. Achtibat, Hatefi, Dreyer, Jain, Wiegand, **Lapuschkin**, Samek (2024).

"AttnLRP: Attention-Aware Layer-wise Relevance Propagation for Transformers".

In: Proceedings of the 41st International Conference on Machine Learning (ICML) 135–168.

In this paper we adapt the popular LRP method to contemporary Transformer architectures, yielding state-of-the-art explanation quality at exceptional computational efficiency. This acievement enables the analysis of the reasoning processes of Vision Transformers and LLMs in real time, eg. in chat bot applications.

2. Achtibat, Dreyer, Eisenbraun, Bosse, Wiegand, Samek and Lapuschkin (2023).

"From attribution maps to human-understandable explanations through Concept Relevance Propagation". In: *Nature Machine Intelligence* 5(9):1006–1019.

A paper introducing the second generation of Explainable Artificial Intelligence with human-readable and abstract concept-based explanations.

3. Pahde, Dreyer, Samek and Lapuschkin (2023).

"Reveal to Revise: An Explainable AI Life Cycle for Iterative Bias Correction of Deep Models".

In: Proceedings of the International Conference on Medical Image Computing and Computer-Assisted Intervention 596–606.

This paper is dedicated to the incorporation of XAI as a standard component into the life cycle of Artificial Intelligence systems, with the intent to improve performance, reliability, and safety of AI.

4. Hedström, Weber, Krakowczyk, Bareeva, Motzkus, Samek, Lapuschkin and Höhne (2023).

"Quantus: An Explainable AI Toolkit for Responsible Evaluation of Neural Network Explanations and Beyond".

In: Journal of Machine Learning Research 24(34):1–11.

In this paper we present the Quantus toolkit, the first-ever comprehensive XAI evaluation toolkit, constituting a well-organized collection of metrics and tutorials for evaluating explainable models, driven by community contributions.

5. Montavon, Lapuschkin, Binder, Samek and Müller (2017).

"Explaining NonLinear Classification Decisions with Deep Taylor Decomposition".

In: Pattern Recognition 65:211–222.

A paper discussing the mathematical foundation of LRP and its properties. Pattern Recognition Best Paper Award and Pattern Recognition Medal winner of 2020.

6. Bach, Binder, Montavon, Klauschen, Müller and Samek (2015).

"On Pixel-wise Explanations for Non-Linear Classifier Decisions by Layer-wise Relevance Propagation". In: *PLoS ONE* 10(7):e0130140.

A very influential and early work on local XAI, introducing the widely used Layer-wise Relevance Propagation method. This work has so far received over 5400 citations as counted by Google Scholar.

Applications of XAI

7. Anders, Weber, Neumann, Samek, Müller and Lapuschkin (2022).

"Finding and Removing Clever Hans: Using Explanation Methods to Debug and Improve Deep Models". In: *Information Fusion* 77:261–295.

The authors' first work in a series dedicated to the exploitation of knowledge derived from XAI for the improvement of performance and robustness of AI systems.

8. Yeom, Seegerer, Lapuschkin, Binder, Wiedemann, Müller and Samek (2021).

"Pruning by Explaining: A Novel Criterion for Deep Neural Network Pruning".

In: Pattern Recognition 115:107899.

In this paper we leverage information about the importance of latent neural network structures obtained through XAI, in order to drastically reduce over-parameterization by identifying and removing non-critical components, resulting in up to 95% smaller models without loss of performance, and thus strong gains in energy and run time efficiency.

9. Horst, Lapuschkin, Samek, Müller and Schöllhorn (2019).

"Explaining the Unique Nature of Individual Gait Patterns with Deep Learning".

In: Scientific Reports 9:2391.

This paper is representative of many, in which we leverage techniques from XAI in domains where model transparency is critical, enabling for the first time the application of more powerful non-linear predictors beyond traditional linear systems in a feasible manner.

10. Lapuschkin, Wäldchen, Binder, Montavon, Samek and Müller (2019).

"Unmasking Clever Hans Predictors and Assessing what Machines Really Learn".

In: Nature Communications 10:1069.

One of the first papers to rigorously perform model- and data analysis through the lens of XAI, adding a voice of caution to the ongoing excitement about machine intelligence.

11. **Lapuschkin**, Binder, Montavon, Müller and Samek (2016).

"Analyzing Classifiers: Fisher Vectors and Deep Neural Networks".

In: Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition (CVPR) 2016:2912-2920. This is the first paper to use XAI to analyze and document differences in the behavior of state of the art predictors from several epochs of AI, in turn illuminating their use of yet unknown confounding features embedded in widely used computer vision benchmark datasets, critically scrutinizing previous key results from AI research.