

SAMYAK JAIN

Pre-Doctoral Researcher, Microsoft Research

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[LinkedIn](#) ◇ [Github](#) ◇ [Webpage](#) ◇ [Google Scholar](#) ◇ [Twitter](#)

EDUCATION

Indian Institute of Technology (BHU) Varanasi

August 2018 - May 2023

Integrated Dual Degree (B.Tech + M.Tech) in Computer Science - CGPA : **9.55**/10

[Master's Thesis](#)

AREAS OF INTEREST

Research topics: AI safety, Science of deep learning, Interpretability, Learning dynamics, Reasoning, Planning

Sub-topics: Adversarial robustness, Red teaming, Safety fine-tuning, Compositional generalization, Phase transitions, Mode connectivity, Domain generalization, Reward hacking, Cooperative alignment, Lottery ticket hypothesis.

EXPERIENCE

Microsoft Research India

July 2024 - Present

Research Fellow

Mentor [Navin Goyal](#)

Project: Developing a better understanding of why lottery tickets exist using tools from interpretability.

Five AI and Torr Vision Group, University of Oxford

October 2023 - June-2024

Research Intern

Mentor [Puneet Dokania](#)

Project: Demonstrated the mechanisms involved behind the success of jailbreaking attacks.

Krueger AI Safety Lab, University of Cambridge

May 2023 - October-2023

Research Intern

Mentor [David Krueger](#)

Project: Showed that fine-tuning learns minimal transformations of a pretrained model's capabilities, like a 'wrapper'.

Vision and AI Lab, Indian Institute of Science, Bangalore

May 2020 - May-2023

Research Intern

Mentor [Venkatesh Babu](#)

Project: Built more effective and efficient adversarial training methods, achieving SOTA performance on leaderboards.

Theoretical Foundations of AI Lab, Technical University of Munich

May 2021 - August-2021

Research Intern

Mentor [Debarghya Ghoshdastidar](#)

Project: Worked on understanding the learning dynamics of linear autoencoders.

PUBLICATIONS

- **What Makes Safety Fine-tuning Methods Safe? A Mechanistic Study**
Samyak Jain, Ekdeep Singh, Kemal Oksuz, Tom Joy, Phil Torr, Amartya Sanyal, Puneet Dokania
ICML workshop on Mechanistic Interpretability, 2024 (**Spotlight**)
NeurIPS 2024 [main code](#)
- **Mechanistically analyzing the effects of fine-tuning on procedurally defined tasks**
Samyak Jain*, Robert Kirk*, Ekdeep Singh*, Hidenori Tanaka, Robert Dick, Tim Rocktaschel, Edward Grefenstette, David Krueger
ICLR 2024 [main code](#)
- **Towards Understanding and Improving Adversarial Robustness of Vision Transformers**
[Samyak Jain](#), Tanima Dutta
CVPR 2024 [main](#)
- **DART: Diversify-Aggregate-Repeat Training Improves Generalization of Neural Networks**
[Samyak Jain*](#), Sravanti Addepalli*, Pawan Sahu, Priyam Dey, RV. Babu
CVPR-2023 [main code](#)
- **Efficient and Effective Augmentation Strategy for Adversarial Training**
Sravanti Addepalli*, [Samyak Jain*](#), RV. Babu
NeurIPS 2022 [main code](#)
- **Scaling Adversarial Training to Large Perturbation Bounds**
Sravanti Addepalli*, [Samyak Jain*](#), Gaurang Sriramanan, RV. Babu
ECCV 2022 [main code](#)
- **Boosting Adversarial Robustness using Feature Level Stochastic Smoothing**
Sravanti Addepalli*, [Samyak Jain*](#), Gaurang Sriramanan*, RV. Babu
SAIAD Workshop CVPR 2021 [main code](#)

FEATURED ACADEMIC PROJECTS AND COLLABORATIONS

Understanding the lottery ticket hypothesis [Navin Goyal](#)

- Discovered that neurons forming lottery tickets have a high projection with the final model at initialization itself.
- Analytically showed that high projection leads to rapid rise in norm of such neurons, leading to faster convergence.

Mechanistic understanding of safety fine-tuning and jailbreaking attacks [Puneet Dokania](#), [Ekdeep Singh](#), [Amartya Sanyal](#), [Phil Torr](#)

- Observed that safety fine-tuning projects unsafe samples into model's null space, thereby leading to safe behavior.
- Demonstrated that the learned projection is low-ranked in nature, which makes it easy to craft jailbreaks.
- [Gemma Scope](#) highlighted that using sparse autoencoders based on insights in this work could help improve safety.

Mechanistic understanding of fine-tuning [Robert Kirk](#), [Ekdeep Singh](#), [David Krueger](#), [Hidenori Tanaka](#), [Tim Rocktaschel](#), [Edward Grefenstette](#)

- Demonstrated that fine-tuning is unable to alter the model mechanistically, but rather gives a pretense of change.
- Proposed reverse fine-tuning to demonstrate this, which has now become a staple method to evaluate unlearning.
- [Follow-up](#) works have used key insights from our work to counter use of safety fine-tuning as an assurance protocol.

Exploring loss basin to find generalized solutions [RV. Babu](#), [Sravanti Addepalli](#)

- Proposed to train diverse models while intermittently averaging their weights to explore the loss landscape.
- Derived upper bounds showing that weight averaging of diverse models in training slows learning of spurious features.
- Proposed method demonstrated improved performance on both in-domain and domain generalization settings.

Using data augmentations effectively in adversarial training [RV. Babu](#), [Sravanti Addepalli](#)

- Showed for the first time that it is possible to use data augmentations effectively in adversarial training.
- Demonstrated that weight space smoothing with single-step attacks can help in preventing catastrophic overfitting.

Aligning adversarial training with Ideal training objectives [RV. Babu](#), [Sravanti Addepalli](#)

- Observed that standard AT methods fail to generalize to larger perturbation bounds due to change in oracle label.
- Proposed a method, which aims to align the model's predictions with the oracle labels of adversarial images.

Understanding gradient masking in vision transformers [Tanima Dutta](#)

- Past works have demonstrated gradient masking in vision transformers, but failed to understand the cause for it.
- Demonstrated that softmax in attention creates floating point errors, which leads to gradient masking in VITs.

SCHOLASTIC ACHIEVEMENTS

- Recipient of the **DAAD WISE** fellowship, a research-oriented scholarship program funded by the German govt.
- Fellow of Berkeley Existential Risk Initiative (**BERI**), which supported my research at University of Cambridge.
- Recipient of Summer Research Fellowship (**SRFP**), which supported my work at Indian Institute of Science.
- All India rank 922 in JEE Advanced 2018 and 346 in JEE Mains 2018 out of 1 million+ candidates.
- Recipient of the KVPY 2018 Fellowship (Indian Institute of Science, Bangalore), given by the Govt. of India.
- Ranked among the **top 300** students in India in the National Olympiads for Maths, Physics, and Astronomy (INMO, INPhO, INAO) in 2018. Ranked **second** in state in National Talent Search Exam (NTSE) 2016.
- Member of [Future of Life-Existential AI Safety Community](#).

INVITED TALKS AND PRESENTATIONS

Mechanistic understanding of safety fine-tuning and jailbreaks ICML mechanistic interpretability workshop.	July 2024
Pitfalls in safety fine-tuning for robust alignment ETH Zurich AI Center.	February 2024
Mechanistic understanding of fine-tuning Krueger AI safety lab, University of Cambridge and Five AI, Oxford.	November 2023

FEATURED POSITIONS AND RELEVANT COURSES

Reviewer: NeurIPS 2024, ICLR 2024, ICML 2023, NeurIPS 2023, CVPR 2023, CVPR 2022, ICLR 2022, NeurIPS 2022.

Outstanding / Highlighted Reviewer Award: NeurIPS 2024, CVPR 2023, CVPR 2022, ICLR 2022

Teaching Assistant: Introduction to Database Management and Introduction to Machine Learning

- Conducted lab classes of undergraduate students with a batch size of over eighty students.
- Worked alongside the professor to design and evaluate lab assignments and final course assessments.

Relevant Courses: Computer Vision (**A**), Applied Deep Learning (**A**), Theory of Computation (**A-**), Artificial Intelligence (**A**), Probability and Stats (**A**), Real Analysis (**A**), Random Processes (**A**), Linear Algebra (**A**), Data Mining (**A**), Computer Graphics (**A***), Calculus (**A**), Signal Processing (**A**), Number Theory (**A-**), Data Structures (**A-**), Algorithms (**A***), Information Security (**A***), Rings and Modules (**A**), Probabilistic Graphical Models and Optimization (online).