

Atharva Naik

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🐙 [atharva-naik](https://github.com/atharva-naik)

🌐 [Atharva Naik](#)



🌐 <https://atharva-naik.github.io/>

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Experience

- 2022 –  **Research Assistant, Carnegie Mellon University** Automated code review generation, Code translation, and Interactive reflection questions for database optimization tutoring.
- 2021  **Research Intern, Technische Universität Darmstadt** Neural Network Architecture for Faithful Interpretability in NLP.
-  **Research Intern, Adobe** RL agent for Creative Human-Human Collaboration.
-  **Research Intern, University of Alberta** Neuro-Symbolic Fuzzy Logic-based Reasoning for Explainable Natural Language Inference.
- 2019-2020  **Student Researcher, Autonomous Ground Vehicle (AGV) Group** Path Planning and Localization for Autonomous Driving.

Education

- 2022 – 2024  **M.S Computer Science, Carnegie Mellon University** in Language Technologies.
GPA: 4.08/4
- 2018 – 2022  **B.Tech. Computer Science, Indian Institute of Technology, Kharagpur**
GPA: 9.66/10

Research Publications

Conference Publications

- 1 **A. Naik**, S. Das, J. Vedurada, and S. Aditya, “Sync: A structurally guided hard negative curriculum for generalizable neural code search,” in *AACL*, 2023.
- 2 Z. Wu, Z. X. Zhang, **A. Naik**, Z. Mei, M. Firdaus, and L. Mou, “Weakly Supervised Explainable Phrasal Reasoning with Neural Fuzzy Logic,” in *ICLR Poster*, 2023.
- 3 Y. Xie, **A. Naik**, D. Fried, and C. Rose, “CMTrans: Improving Code Translation with Comparable Corpora and Multiple References,” in *EMNLP Findings*, 2023.
- 4 S. Bv, J. A. Patel, **A. Naik**, Y. Butala, S. Sharma, and N. Chhaya, “Towards Enabling Synchronous Digital Creative Collaboration: Codifying Conflicts in Co-Coloring,” in *CHI Extended Abstracts*, 2022.
- 5 B. Santra, S. Roychowdhury, A. Mandal, *et al.*, “Representation Learning for Conversational Data using Discourse Mutual Information Maximization,” in *NAACL*, 2022.
- 6 Y. Wang, S. Mishra, P. Alipoormolabashi, *et al.*, “Super-NaturalInstructions: Generalization via Declarative Instructions on 1600+ NLP Tasks,” in *EMNLP*, 2022.
- 7 R. Mukherjee, **A. Naik**, S. Poddar, S. Dasgupta, and N. Ganguly, “Understanding the Role of Affect Dimensions in Detecting Emotions from Tweets: A Multi-task Approach,” in *SIGIR*, 2021.

Preprints

- 1 A. Rao, S. Vashistha, **A. Naik**, S. Aditya, and M. Choudhury, “Tricking LLMs into Disobedience: Understanding, Analyzing, and Preventing Jailbreaks,” 2023. arXiv: 2305.14965 [cs.CL].

Skills

Coding	Python (expert), C/C++, Bash (familiar), Javascript (novice)
Frameworks	PyTorch, HuggingFace, Fairseq, NLTK, spaCy, Tensorflow, FastAPI, Flask, Django, PyQt5, Jupyterlab, OpenCV, Git

Awards

2022	2nd place at Deep Learning Labs OpenAI GPT-3 Hackathon
2021	DAAD WISE Scholarship
	MITACS Globalink Scholarship
	Bronze at Inter IIT Technology Meet: Member of IIT Kharagpur contingent
2019	2nd place at Intelligent Ground Vehicle Competition (IGVC)
2018	All India Rank 1248 in JEE Advanced and 1618 in JEE mains among 1M candidates
	Kishore Vaigynaik Protsahan Yojana (KVPY) Scholarship

Other Projects

Interactive Reflection Prompts for Collaborative Programming in Education (Ongoing)

- Used LLMs to generate alternative solutions for teaching database optimization using SQL.
- Explored the educational value of strategic and contextual interactive reflection questions in a collaborative SQL programming activity with 35 students through pre/post tests.

Reference Free Evaluation of Automated Code Review Generation (Ongoing)

- Revisited evaluation and modeling assumptions for automated code review generation.
- Proposed pragmatics-inspired, model-based reference-free evaluation metrics to measure relevance, informativeness, correctness, and readability of reviews.
- Leveraged contrastive learning to model relevance and abstractness lexicons for informativeness.
- Proposed metrics align better with human judgment compared to current metrics like BLEU

Multilingual Multimodal Contrastive Learning for Unified Representations (Ongoing)

- Using multimodal, multilingual contrastive learning to learn unified representations for code and multiple natural languages for jointly modeling multilingual code generation and translation.
- Data augmentation for code using semantics preserving syntax level transformations

NBQA: Retrieval Augmented Jupyter Notebook Completion

- Leveraged prompting-based query rewriting and semantic code retrieval to augment code generation for Jupyter notebooks.
- Achieved improvements in next-cell code prediction over CodeBLEU and API call coverage