

# Santhosh Sankar

sankar.s@northeastern.edu | linkedin.com/in/santhosh-sankar | 857-277-9864  
santhosh-sankar.github.io/portfolio.github.io | github.com/Santhosh-Sankar

## EDUCATION

### Master of Science in Robotics

Northeastern University, Boston, MA

May 2023

CGPA: 3.838

**Coursework:** Deep Learning, Reinforcement Learning and Sequential Decision Making, Pattern Recognition and Computer Vision, Mobile Robotics, Robot Sensing and Navigation, Robot Science and Systems

### Bachelor of Engineering in Mechanical Engineering

Anna University, Chennai, India

May 2020

## TECHNICAL SKILLS

**Programming and Query Languages:** C++, Python, MATLAB, SQL

**Machine Learning and Parallel Programming:** TensorFlow, PyTorch, Amazon SageMaker, CUDA

**Software Libraries:** OpenCV, PCL, NumPy, SciPy, Matplotlib, pandas, scikit-learn

**Software Tools and Operating Systems:** AWS, ROS, Git, Docker, Nsight Systems, Linux, Windows

## PROJECTS

### Machine Translation with Transformer using AWS

Oct 2023 - Nov 2023

- Built a Transformer-based sequence-to-sequence model to translate German sentences to English using **Python** and **PyTorch**, with a custom module to compute multi-head attention.
- Trained, deployed, and assessed the Transformer model with the Multi30K dataset stored in **Amazon S3** using **Amazon SageMaker** and achieved a high BLEU score of **78%** on the test dataset.

### Chatbot using sequence-to-sequence Transformer

July 2023 - August 2023

- Designed an end-to-end chatbot based on sequence-to-sequence Transformer with **Python** and **TensorFlow** and implemented a data preprocessing pipeline to extract and process the input data for training.
- Trained the model with Kaggle TPUs on Cornell movie dialog corpus with over 200,000 conversations and improved the BLEU score by **3%** with a word tokenizer after significant hyperparameter tuning.

### Parallel image processing with CUDA

May 2023 - Jun 2023

- Developed a baseline code with **OpenCV** in **C++** for processing input images and frames from the live video feed with Gaussian, Sobel, magnitude, quantization, and cartoonization filters.
- Parallelized the sequential baseline code with **CUDA** and profiled it with **Nsight Systems**, optimizing memory access and accelerating performance by at least **75%**, relative to the baseline.

### Siamese networks with attention for large-scale landmark retrieval

Mar 2023 - Apr 2023

- Devised a data preprocessing pipeline in **Python** that yielded new landmark pairs and similarity scores from 1.6 million images in the Google Landmark Dataset (GLDv2) while training to prevent overfitting.
- Designed and trained Siamese networks in **TensorFlow** on GPU cluster, pairing ResNet-101 with spatial, channel, CBAM, and SE attention modules as subnetworks to retrieve images similar to a given image.
- Achieved **9.46%** higher mean Average Precision (mAP) than the baseline classifier with the SE module and improved the mAP further by **4.4%** with an ensemble of spatial, SE, and CBAM attention modules.

### Pedestrian tracker and counter using YOLOv3 and DeepSORT

Oct 2022 - Dec 2022

- Constructed YOLOv3 with **Python** and **TensorFlow** utilizing the Darknet-53 architecture to perform pedestrian detection on video frames and incorporated weights pre-trained on the COCO dataset.
- Integrated the YOLOv3 object detector with the DeepSORT algorithm for pedestrian tracking and counting and visualized the pedestrian paths over the most recent 50 frames with a **Python** script.
- Evaluated DeepSORT with YOLOv3 and Faster RCNN on MOT16 benchmark, with YOLOv3 attaining **6%** higher tracking accuracy (MOTA) and **3%** higher tracking precision (MOTP) over Faster RCNN.

### Performance comparison of RL algorithms in Super Mario Bros

Oct 2022 - Dec 2022

- Developed and trained DQN, DDQN, and PPO deep reinforcement learning agents to complete one level of Super Mario Bros using **Python** and **PyTorch** and assessed the performance over 1000 episodes.
- Achieved **1%** and **30%** higher success rates, **2.4%** and **22.7%** higher average returns, and **4.1%** and **34.9%** lower mean steps per episode with DDQN and PPO agents respectively, over the DQN agent.