

# Nishant Gurunath

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## OBJECTIVE

Seeking employment to develop algorithms and software for machine learning

## CORPORATE ROLES

### TEXAS INSTRUMENTS

DESIGN ENGINEER, 2016-2018

## EDUCATION

### CARNEGIE MELLON UNIVERSITY (CMU)

MASTER OF SCIENCE IN ELECTRICAL AND COMPUTER ENGINEERING

Expected December 2019

Pittsburgh, PA

GPA: 3.78 / 4

### INDIAN INSTITUTE OF TECHNOLOGY, BOMBAY (IIT BOMBAY)

BACHELOR OF TECHNOLOGY +

MASTER OF TECHNOLOGY IN

ELECTRICAL ENGINEERING +

MINOR IN COMPUTER SCIENCE

Graduated June 2016

Mumbai, India

GPA: 8.57 / 10

## COURSEWORK

### CMU

Introduction to Machine Learning

Introduction to Deep Learning

Foundations of Computer Systems

Probabilistic Graphical Models

Computer Vision

How to Write Fast Codes

Machine Learning for Signal

Processing

### IIT BOMBAY

Data Structures and Algorithms

Probability and Random Processes

Data Analysis and Interpretation

Linear Algebra

Graph Theory

## SKILLS

### PROGRAMMING

Python • C/C++ • HDL • SQL

### PACKAGES

Pytorch • MATLAB • GDB • Falcon

## WORK EXPERIENCE

### MACHINE LEARNING INTERN | MOODY'S ANALYTICS | NEW YORK, NY

May 2019 – August 2019 | Return Offer as Associate Director: Machine Learning

- Initiated a **new product** that aimed to extract structured information from text data
- Implemented a workflow around AllenNLP OIE and SRL models to obtain relationship knowledge graph among named entities - people, organizations and locations
- Performed Text Normalization, Entity Disambiguation, Named Entity Recognition, Coreference Resolution and Neo4j graph database queries as part of the workflow
- Created APIs for two product lines and customized Framenet based argument parsing

### RESEARCH ASSISTANT - NLP | SEPARABL: DISENTANGLEMENT IN SPEECH | CMU

January 2019 - Present | Prof. Alan Black | LTI

Submitted in AAAI 2020 | [Paper](#) | [Samples](#) | Currently Working on its Product

- **Proved** that multinode VAE can be used to separate speech and music in audio
- Experimentally determined the number of latent nodes required for source separation; showed that the same can be determined from input data distribution
- Established improvement in Speech Synthesis performance using separated speech

## ACADEMIC PROJECTS

### HIERARCHICAL REINFORCEMENT LEARNING | CMU

Spring 2019 | Probabilistic Graphical Models | Prof. Eric Xing

- Proposed to learn policies simultaneously for two agents, manager and worker, working at different temporal scales to target environments with sparse rewards
- Demonstrated that method is applicable with all reinforcement learning algorithms
- Achieved a **60%** improvement in reward for the Stochastic MDP environment
- **Conquered** sparse reward robotic environments - ReacherV2 and FetchReachV2 where standard/ flat reinforcement learning methods fail completely

### IDENTIFYING DUPLICATE QUESTIONS ON QUORA | CMU

Fall 2018 | Introduction to Machine Learning | Prof. Ziv Bar-Joseph

- Designed Siamese models over Glove embedding to detect semantics of the questions
- Modeled a CNN with parallel kernels to capture meaning from different parts of the question; improved on the CNN based state-of-the-art accuracy by **1.5 percent points**
- Obtained a further 2 percent point improvement over CNN with BiLSTM model

### AUDIO BASED MULTIMEDIA EVENT DETECTION AND BAYESIAN LEARNING | CMU

Fall 2018 | Introduction to Deep Learning | Prof. Bhiksha Raj

- Designed ResNet based event classification model on limited training data (YLI-MED)
- Developed data augmentation method to enhance dataset by temporal perturbation
- Applied Bayesian learning on network parameters to add robustness with regard to unknown data; achieved **4 percent point** improvement over the baseline CNN model

### LISTEN, ATTEND AND SPELL: ATTENTION MODEL | CMU

Fall 2018 | Introduction to Deep Learning | Prof. Bhiksha Raj

- Created a speech to text generation LAS model using cascaded LSTM networks
- Designed a pyramidal BiLSTM speech encoder to reduce computational complexity
- Modeled an attention based LSTM transducer which generates a distribution over the next character conditioned on all previous characters; Obtained an accuracy of 75%

### SPEAKER CLASSIFICATION AND VERIFICATION: TRANSFER LEARNING | CMU

Fall 2018 | Introduction to Deep Learning | Prof. Bhiksha Raj

- Created a speech classification model using a CNN with residual blocks
- Trained the network to learn speaker embeddings during the classification task
- Used the learned embeddings to obtain similarity between a given speaker pair
- Obtained an EER  $\approx 5$  for a test sample consisting 10000 speaker pairs