# HARSHVEER SINGH

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## **EDUCATION**

#### Thapar Institute of Engineering and Technology(TIET)

Iuly '17 - Present

Bachelor of Engineering (B.E.) in Electronics and Computer Engineering

## **EXPERIENCE & TECHNICAL PROJECTS**

#### Academic Head, ACM Student Chapter, T.I.E.T.

July '19 - July '20

Planned the academic curriculum: topics, number of classes, mentor assignment. Mentored the Machine Learning research group
Mentored teams for PECFEST Hackathon '19. SATURNALIA Hackathon '19

Hentored teams for Fee Est Flackathon 13, SATOMALIA Flackathon 13

Tinotifications Dec '18

Developed an event management and publicity platform for TIET, (currently in use, exclusive for TIET students), backboned with a Transformer based attention model (Poly Encoders), to pick next favorable task to operate on, ensuring smooth functioning of a given student organization at TIET.

Hackathons 2017-2018

SATURNALIA Hackathon '17, used bayesian prior to regress rural health service data.

PECFEST Hackathon '17, used VAEs for clustering supply chain records of a given organization. Jenson-Shannon divergence metric performed better for estimating the posterior than basic KLdivergence.

PECFEST Hackathon '18, developed a question answering system based on gpt2, to help under priviledged students to learn subjects on their own.

#### RESEARCH EXPERIENCE

#### Adversarial training for Facebook's blender bot

June'20 - Present

Aiming at a self play regime for conversational agents, and extending that to competitive conversation where an agent discriminates the output distribution of the other agent against human dialogue distribution.

#### Poly encoder regime for fine-tuning decoder-only model (GPT-2)

May' 20

Decoder model fine tuned like such, apparently is more robust to inductive bias than encoder model even though encoder reached better recall@k/C score.

#### Analytical study of success of Batch Norm

Nov'19 - Dec'19

Showed that batch normalization smooths the loss surface and how it brings that effect, through the study of eigenvalues of the hessian of input matrix.

## Beta2 variation regime for Adam Optimizer

May '18 - July '18

Developed a novel regime for varying beta2 hyper-parameter of Adam (adaptive momentum), preventing Adam from getting stuck in sub-optimal minima. A similar result was also shown in the parallel but completely independent research Sashank I. Reddi et al.

## **SKILLS**

**Mathematics**: Linear algebra, Statistics, Bayesian Inference, Probabilistic modeling **Programming**: Data structures and algorithms, Python, PyTorch, C, C++, HuggingFace, NumPy, TensorFlow, Keras, scikit- learn, pandas, matplotlib. **Web**: NodeJS, ReactJS, VanillaJS, MongoDB, Nginx, Multiple Third-party APIs. **Utilities**: Git/GitHub, Linux(Kernel Programming), Atom, SublimeText, LaTeXs

### **PUBLICATIONS**

#### The real reason why BatchNorm works

https://towardsdatascience.com/why-batchnorm-works-518bb004bc58

Code for Poly-encoder got featured at the official10 paperswithcode.com profile of Poly-encoders[Humeau et al.], Code - <a href="IStringII/Poly-encoders">IStringII/Poly-encoders</a>

## TALKS AND PRESENTATIONS

| Causality and its importance in variational inference and EM, TIET  | Jan '20 |
|---|---------|
| Inductive bias in machine learning models, TIET                     | Oct '19 |
| Effect of constraining the posterior to a Gaussian in VAEs, PECFEST | Nov '17 |

## **CURRENT WORK**

Evolution of ML algorithms from scratch using only basic mathematical operations as building blocks, applying evolutionary methods to automatically find the code for complete ML algorithms.

Finally aiming at creating such a regime of environment and agents, where agents can learn to code their environment and eventually act as an evolved drop-in replacement of their environment, hence bootstrapping the process. Inspired by neural architecture search and AutoML-Zero by GoogleAI.

Analytical study of GPT3 and other predecessors, to figure out why and how they show signs of reasoning, and prove their reasoning to be not falsifiable.

Why deep attention models achieve better adaptability to open domain tasks. An opposite perspective to transfer learning.

Removing redundancy in GPT3, eventually making it work with minimal computation.