Brian Lester

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Education

University of Pittsburgh   *Pittsburgh, PA*
DOUBLE MAJOR, 3.51. COMPUTER SCIENCE WITH HONORS, 3.86. NEUROSCIENCE, 3.19.   *2012–2016*
Shanghai American School   *Shanghai, China*
HIGH SCHOOL   *2008–2012*
Advanced Placement Curriculum. Team lead for FIRST Robotics Competition.

Selected Publications   h-index: 6

The Power of Scale for Parameter-Efficient Prompt Tuning
BRIAN LESTER, RAMI AL-RFOU, NOAH CONSTANT
EMNLP 2021, Online. 273 citations

Finetuned Language Models Are Zero-Shot Learners
JASON WEI, MAARTEN BOSMA, VINCENT Y ZHAO, KELVIN GUU, ADAMS WEI YU, BRIAN LESTER, NAN DU, ANDREW M DAI, QUOC V LE
ICLR 2022, Online. 136 citations

SPoT: Better Frozen Model Adaptation through Soft Prompt Transfer
TU VU, BRIAN LESTER, NOAH CONSTANT, RAMI AL-RFOU, DANIEL MATTHEW CER
ACL 2022, Dublin, Ireland and Online. 23 citations

An Effective Label Noise Model for DNN Text Classification
ISHAN JINDAL, DANIEL PRESSEL, BRIAN LESTER, MATTHEW NOKLEBY
NAACL 2019, Minneapolis, Minnesota. 21 citations

A Library for Rapid Modeling, Experimentation and Development of Deep Learning Algorithms targeting NLP
DANIEL PRESSEL, SAGNIK RAY CHOUDHURY, BRIAN LESTER, YANJIE ZHAO, MATT BARTA
ACL 2018; *NLP-OSS Workshop*, Melbourne, Australia. 12 citations

Multiple Word Embeddings for Increased Diversity of Representation
BRIAN LESTER, DANIEL PRESSEL, AMY HEMMETER, SAGNIK RAY CHOUDHURY, SRINIVAS BANGALORE
Preprint arXiv:2009.14394, 2020. 5 citations

Constrained Decoding for Computationally Efficient Named Entity Recognition Taggers
BRIAN LESTER, DANIEL PRESSEL, AMY HEMMETER, SAGNIK RAY CHOUDHURY, SRINIVAS BANGALORE
EMNLP 2020; *Findings*, Online. 3 citations

iobes: Library for Span Level Processing
BRIAN LESTER
ACL 2020; *NLP-OSS Workshop*, Online. 2 citations

Dynamically Adjusting a Voice Recognition System
BRIAN LESTER, SORIN M PANAINTE
US Patent 9,984,688, 2018. 1 citation

Baseline: Strong, Extensible, Reproducible, Deep Learning Baselines for NLP
DANIEL PRESSEL, BRIAN LESTER, SAGNIK RAY CHOUDHURY, MATT BARTA, YANJIE ZHAO, AMY HEMMETER
NuerIPS 2018; *OSS Workshop*, Montreal Quebec. 1 citation

Intent Features for Rich Natural Language Understanding
BRIAN LESTER, SAGNIK RAY CHOUDHURY, RASHMI PRASAD, SRINIVAS BANGALORE
NAACL 2021; *Industry Track*, Online.

Overcoming Catastrophic Forgetting in Zero-Shot Cross-Lingual Generation
TU VU, ADITYA BARUA, BRIAN LESTER, DANIEL MATTHEW CER, MOHIT IYYER, NOAH CONSTANT
Preprint arXiv:2205.12647, 2022.

Work Experience

Google Brain   *Mountain View, California*
SENIOR RESEARCH ENGINEER   *2020–Present*
Deep Learning research with a focus on Natural Language Processing, large pre-trained models, and zero-shot transfer.
- Prompt Tuning: an efficient method of controlling large frozen pre-trained language models based on T5. Matches performance of full fine-tuning using only 0.003% of the parameters. Open-sourced of our codebase and it has enabled 3 published papers, 1 product launch, and at least 5 in-flight papers.
- Flan: Multitask training for a 137 billion parameter transformer-based decoder-only language model to create a model that is more effective at zero-shot prompting and performs better using Prompt Tuning.
- SPoT: Using multitask prompts as strong initialization for Prompt Tuning resulting in increased performance. Also used prompt similarity to estimate task similarity and to predict transferability.
- Added partial network training, lazy loading, and pre-filling of the auto-regressive cache to t5x, the open-source reimplementation of T5 in Jax. This final change reduced inference latency from 30 seconds to 2.4.

Interactions
MACHINE LEARNING ENGINEER

Ann Arbor, Michigan
2018–2020

Built production grade deep learning solutions and lead research efforts to push the boundaries of performance.
- Designed novel neural network architectures for calibrated intent detection, slot filling, and named entity linking using ConvNets, bLSTM-CRFs, ranking models, and transformer-based seq2seq models.
- Designed label space, annotation guidelines, and data collection method for NLU component of dialogue systems.
- Created a cloud-native model training platform based on declarative pipelines and kubernetes. Built a deployment platform that powers NLU for multiple production dialogue systems.
- Built efficient, batched implementations of complex neural network architectures such as Beam Search. My CRF implementation reduced training time by a factor of 10.

Trove
LEAD MACHINE LEARNING RESEARCH ENGINEER

Ann Arbor, Michigan
2017–2018

Created a model training and serving platform that processed 200 million emails per day. Provided technical leadership to the ML team.
- Designed ConvNets for text classification to find sentences that contain questions. This powered a user-facing feature and was used to featurize the social graph created from email.
- Created neural ranking model was used to find coreferent mentions in the text and provide context to users.
- Used lexical features, as well as connectivity information in the email social graph, to identified bot accounts.

Visteon Corporation
SOFTWARE ENGINEERING INTERN

Van Buren, Michigan
2015

Designed an adaptive system to minimize voice recognition errors based on ASR confidence scores. We patented this system and it is used in Mazda Cars.

Presentations

Prompt Tuning

University of Michigan
University of North Carolina
2021–2022

DEEP LEARNING

An overview of my work on Prompt Tuning, as well as our work—Flan and SPoT—directly built on Prompt Tuning. The talk includes a collection of insights about the behavior of soft prompts aggregated from others' followup work.

NeurIPS Spotlight Talk on Mead-Baseline
DEEP LEARNING

NeurIPS OSS Workshop
December 2018

A spotlight talk, at the Open Source Software workshop at NeurIPS 2018, about our open-source toolkit, Mead-Baseline.

Confidence and Calibration of Neural Network Models
DEEP LEARNING

EMU ML Conference
March 2020 [1]

An overview of techniques used to adjust model calibration, evaluation of models that have the ability to "reject" decision with low confidence, and their uses in the NLU unit of a production dialogue system.

Padding in Neural Networks for Natural Language Processing
NATURAL LANGUAGE PROCESSING

A2D-NLP
February 2020

A survey of NLP building blocks with a focus on correctness and the need for padding in complex situations as well as places it is unexpected, like max-pooling following a 1D convolution.

Optimization via NumPy and Cython
NUMERICAL COMPUTATION

Michigan Python Meetup
January 2020

I use a series of optimizations for computing pairwise Manhattan distance to introduce core NumPy concepts and Cython to reduce the runtime from multiple hours to just seconds.

Input Representations of Deep Neural Networks
DEEP LEARNING

PyData Ann Arbor
October 2017

Using learned character-compositional input representations to create Deep Neural Networks with an open vocbaulary.

Skills

Deep Learning	Extensive experience building novel Neural Network architectures, generally for NLP. High-performance training with Data and Model Parallelism, including multihost distributed training on TPU.
Infrastructure	Build and deploy with Kubernetes, Docker, Flux, MongoDB, Apache Nifi, Github Actions, and GitLab CI/CD. Experience training large neural networks on Google Cloud (GCP)
Toolkits	Jax, Flax, PyTorch, Tensorflow, NumPy, Pandas, Faiss, SpaCy, NLTK, Tensorflow-Datasets, Seaborn, and Matplotlib.
Languages	Python, Cython, Java, C, Javascript, C++, Elisp, and LaTeX.

Service & Public Scholarship

REVIEWING

2022	IEEE Transactions on Affective Computing
2022	ARR: ACL Rolling Review
2022	NAACL
2020–2021	Computer Speech and Language
2019	CoNLL

PUBLIC SCHOLARSHIP

2020 [1]	**a²-dlearn**: Helped organize logistics, recruit speakers, and acquire funding through sponsorships

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[1]Canceled due to COVID-19