# ARUSHI JAIN

Mila, McGill University, Montreal, Canada

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#### **RESEARCH INTERESTS**

Reinforcement Learning (RL), Safe RL, Off-Policy RL, Constrained MDPs.

#### EDUCATION

Ph.D., Computer ScienceGPA: 4.0MSc, Computer ScienceGPA: 4.0	
MSC, Computer Science GPA: 4.0	
7.611	00/4.00 Sept 2017 - Sept 2019
Mila	
Supervisor: Doina Precup, <u>Pierre-Luc Bacon</u>	

Bachelor of Technology, Computer Science and Engineering GPA: 9.42/10.00 Aug 2012 - May 2016 Image Analysis and Biometrics (IAB) lab Supervisor: Mayank Vatsa, Richa Singh

## AWARDS

• Bourse du Centre de Recherche Informatique de Montréal (CRIM) pour Études Supérieures (2018-2019) Graduate Scholarship provides financial support to only one master and a doctoral student.

## PUBLICATIONS

- 1. Variance Penalized On-Policy and Off-Policy Actor-Critic[Paper][Poster][Slide] Arushi Jain, Gandharv Patil, Ayush Jain, Khimya Khetarpal, Doina Precup Accepted to the Thirty-Fifth AAAI Conference on Artificial Intelligence (AAAI), 2021.
- Safe Option-Critic: Learning Safety in the Option-Critic Architecture [Paper] [Poster] [Slide] Arushi Jain, Khimya Khetarpal, and Doina Precup Accepted in Adaptive Learning Agents (ALA), ICML Workshop, 2018. Accepted in The Knowledge Engineering Review (KER) Journal 2021. [Impact Factor 1.07]
- Safe Actor-Critic[Paper][Poster][Slide] Arushi Jain, Ayush Jain, and Doina Precup Accepted in Safety, Risk and Uncertainty in RL, Uncertainty in AI (UAI) Workshop, 2018. Accepted in Women in ML (WiML), (NeurIPS) Workshop, 2018.
- 4. Learning Options using Constrained Return Variance[Paper][Poster] Arushi Jain and Doina Precup

Accepted in Safety and Robustness in Decision Making, NeurIPS Workshop 2019.

5. Safe Policy Learning with Constrained Return Variance[Paper][Poster] Arushi Jain

Accepted in Graduate Student AI Symposium held at **Canadian AI Conference** 2019. The proceeding is also published in the **LNAI** Series by Springer Verlag.

- Safe Hierarchical Policy Optimization using Constrained Return Variance in Options[Paper] Arushi Jain, Doina Precup Accepted at RLDM 2019.
- Safety using constraint variance in policy-gradient methods[Thesis] Arushi Jain Master Thesis, McGill University, Montreal, March 2020.

## **RESEARCH & WORK EXPERIENCE**

## Mila, RL Lab, McGill University

Graduate Student advised by Prof. Doina Precup, Pierre-Luc Bacon

- Real-world tasks with different competing objectives or constraints, it is not always feasible to evaluate the performance of a given policy/behavior. For example, one wants to assess the performance of a given policy in recommendation systems. However, allowing a potentially bad policy that does not abide system's constraints to interact with customers can lead to unpredictable losses. To overcome this problem, we propose to predict by learning from a different **off-policy** that follows constraints and also leads to smaller variability in the performance than the standard on-policy approach.
- We designed a novel and generic **on-policy and off-policy safe actor-critic** framework to learn safe policies such that objective function trades off between minimizing the effects of model uncertainty and maximizing the mean return.
- We worked on introducing novel safety solutions in Options (hierarchical RL) by deriving a policy-gradient algorithm where variance in the temporal difference (TD) error is regularized.

## Mila, Amii

## Mentored by Sharan Vaswani, Reza Babanezhad

We study planning in infinite horizon Constrained MDPs, where we use parameter-free algorithm from online linear optimization to learn both the policy and follow the constraints. Unlike gradient descentascent methods, it does not require parameter tuning.

## SPORTLOGiQ, Montreal, Canada

## RL Research Intern mentored by Norm Ferns

Worked on formally comparing and contrasting the agents in a given Markov Decision Process (MDP) by coming up with behavioral pseudo-metrics following the work on *lax-bisimulation*.

## Borealis AI, Edmonton, Canada

AI Research Intern mentored by Nidhi Hedge

Worked on **safe recommendations** in RL setting which would help provide recommendations with certain guarantees. This would provide life-long learning as well as long-term goal maximization which would be particularly useful in financial applications.

## Microsoft Research (MSR), India

Research Fellow advised by Sundararajan Sellamanickam, Arun Iyer

• Developed a **service monitoring and diagnostic tool** for unsupervised hierarchical monitoring of services which replaced a heuristic-based system failing to jointly model multivariate time-series.

• Worked on **entity matching problem across heterogeneous sources** with missing data in the source.

• Worked on **email solutions using DNN** to build intelligent approaches to solve problems like multi-label classification, learning a useful representation of email content, etc.

September 2017 - Present

August 2021 - Present

June 2019 - Sept 2019

May 2018 - Aug 2018

June 2016 - July 2017

## **TECHNICAL SKILLS & COURSEWORK**

Programming Languages: Python, Tensorflow, Pytorch, C

**Coursework**: Reinforcement Learning, Probabilistic Analysis of Algorithms, Applied ML, Theoretical ML, Probabilistic Graphical Modeling, Matrix Computation, Mathematical Foundation of ML, Reinforcement Learning and Optimal Control.