The 8th Cluster Workshop on Intelligent Systems and Machine Learning

March 17 (Friday), 2023

4:00pm-6:00pm (Alberta Time)

Organized by: AU FST ISML Research Cluster

Microsoft Teams meeting

Join on your computer, mobile app or room device

Click here to join the meeting

Meeting ID: 214 259 392 930

Passcode: DEsBbV

Download Teams | Join on the web

Or call in (audio only)

<u>+1 647-749-1732,,718570990#</u> Canada, Toronto

(866) 598-0754,,718570990# Canada (Toll-free)

Phone Conference ID: 718 570 990#

Keynote Presentation

Keynote Presenter	Prof. Liang Chen (University of Northern British Columbia, Canada)
Title	From Computer's Point of View
Abstract	What can Computing Machines do better than human beings? The jobs that require "massive" computing resources! But what are we going to "teach" a machine if we human beings do not know how to do a job? There is a gap between our expectation on computing performances and our knowledge on how to do a job. This talk argues that the objective of high accuracy in related research for such jobs lead to nowhere but a general "trial and error" process; therefore "stability" rather than "accuracy" should be applied as the evaluation standard in developing AI algorithms.

Presenter's	Liang Chen carried a bacheror 5 degree in comparer software at Hadzhong
short bio	University of Science and Technology in 1988 and a doctorate in computer
	science from the Software Institute at Chinese Academy of Sciences in 1994.
	Since then, he had worked at various universities and companies in China,
	Japan and France. He has been with the Department of Computer Science at the
	University of Northern British Columbia since 2001 where he is now a
	professor. His research interests include pattern recognition, image processing,
	computational geometry, intelligent language tutoring system, data mining, and
	bioinformatics.

Invited Presentation

Abstract Associative art classific accuracy per classification	le Ensemble Models For Associative Classification e classifiers have shown competitive performance with state-of-the- cation methods for predicting class labels. In addition to their erformance, associative classifiers produce human-readable rules for on, providing an easier way to understand their decision-making owever, the presence of high-dimensional feature vectors in the
Abstract Associative art classific accuracy per classification	e classifiers have shown competitive performance with state-of-the- cation methods for predicting class labels. In addition to their erformance, associative classifiers produce human-readable rules for on, providing an easier way to understand their decision-making
Abstract Associative art classific accuracy per classification	e classifiers have shown competitive performance with state-of-the- cation methods for predicting class labels. In addition to their erformance, associative classifiers produce human-readable rules for on, providing an easier way to understand their decision-making
art classific accuracy po	cation methods for predicting class labels. In addition to their erformance, associative classifiers produce human-readable rules for on, providing an easier way to understand their decision-making
accuracy po	erformance, associative classifiers produce human-readable rules for on, providing an easier way to understand their decision-making
classification	on, providing an easier way to understand their decision-making
	•
process. Ho	owever, the presence of high-dimensional feature vectors in the
Processin	, 1
_	ta impacts these models' performances. In this research, we propose
	le model, Dynamic Ensemble Associative Learning (DEAL), where
I I	ociative classifiers as base learners on feature sub-spaces. We also
1 1	lynamic feature sampling procedure that automatically defines the
	base learners and ensures diversity and completeness among the
	atures. As each of the base learners of our ensemble model is
	le, we can collect the rules from the base learners and examine them
	nd the decision process. Although, understanding this decision
1 -	be complicated if there is a large number of base learners. To that
The state of the s	roduce another ensemble model, Classification by Frequent
	n Rules (CFAR), where we make the decision process of the model
	nderstand. In CFAR, instead of using a voting method for the final
	the ensemble, we rank the generated rules based on their prevalence a subset of the predominant rules for predicting class labels. We use
	s from the UCI repository to evaluate the performance of our
	nodels. Both our ensemble approaches, DEAL and CFAR, eliminate
	emory and runtime requirements of recent associative classifiers,
	for training datasets with high-dimensional feature vectors, all
1	pardizing classification accuracy. CFAR enhances the model's
	ility by removing noisy rules while achieving increased classification
1 -	n most cases in our experiments. While the decision process of DEAL
	nterpretable, CFAR enhances this further by collecting and ranking
	enerated by its base learners to predict the final class label.
Presenter's Md Rayhai	n Kabir completed his master's from the Department of Computing
	niversity of Alberta in 2022. His research interest is on ensemble

models, Explainable AI and Reinforcement Learning. Currently he is working with Dr. Oscar Lin and Dr Ali Dewan.