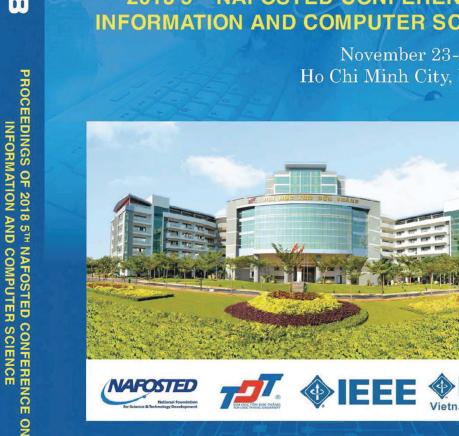
<u>8</u>07

Proceedings of 2018 5TH NAFOSTED CONFERENCE ON **INFORMATION AND COMPUTER SCIENCE**

November 23-24, 2018 Ho Chi Minh City, Vietnam













NICS 2018

PROCEEDINGS OF 2018 5TH NAFOSTED CONFERENCE ON **INFORMATION AND COMPUTER SCIENCE**

Co-organizers





Patrons







Sponsors





NHÀ XUẤT BẢN KHOA HỌC VÀ KỸ THUẬT Địa chỉ: Số 70 Trần Hưng Đạo - Hoàn Kiếm - Hà Nội Diên thoai: 024.39423172 - 024.38220686 Fax: 84.24.38220658 Email: nxbkhkt@hn.vnn.vn Website: http://www.nxbkhkt.com.vn







Table of Contents

Message from the NICS'	18 General and TPC Chairs	X
Executive Committee		xvi
Technical Program Com	mittee	xiz
Additional Reviewers		xxiv
Keynote Abstracts		XXV
•		
Communications ar	nd Natworking	
		T
	nalysis of a Decode-and-Forward Scheme under Physical General Fading Channels	Layer
•	nam (Ho Chi Minh City University of Technology and Education,	
* *	an Phu Tuan (University of Ulsan, Korea), Sol Park (University of Ul	
Korea), Hyu	ng-Yun Kong (University of Ulsan, Korea)	1
Time-Frequency based Kernel	Distribution for Undersampled Non-stationary Signals using	Chirp-
Technology, (Great Britai University o	ng Nguyen (The University of Danang & University of Science and Vietnam), Desmond McLernon (The University of Leeds, United Kinn)), Mounir Ghogho (International University of Rabat, Morocco & Leeds, United Kingdom (Great Britain)), Syed Ali Raza Zaidi (University of Kingdom (Great Britain))	C
	l based on Optimal Capacity for Energy Harvesting Amplifying Networks	
University, I Electronics I Zdralek (Fac University o Electrical En	y (Faculty of Electrical and Electronics Engineering, Ton Duc Thang HCMC, Vietnam, Vietnam), Dac-Binh Ha (Faculty of Electrical and Engineering, Duy Tan University, Da Nang, Vietnam, Vietnam), Jaros Julty of Electrical Engineering and Computer Science, VSB - Technical Ostrava, Ostrava, Czechia, Czech Republic), Miroslav Voznak (Faculti and Computer Science, VSB - Technical University of Ostrava, Czech Republic)	slav al ulty of
complicated sl		
*	Hanoi University of Science and Technology, Vietnam), Khanh-Van noi University of Science and Technology, Vietnam)	17
•	Mode Converter Using Two Silicon Y-Junction Couplers for iplexing Network	
Sciences, Vi Technology, Technology	Linh (Danang University of Science and Technology & Hue Universetnam), Duong Duy (Posts and Telecommunications Institute of Vietnam), Cao Dung Truong (Posts and Telecommunications Institut & Faculty of Electronics Engineering, Vietnam), Hung Nguyen (The	sity of te of
University of	f Danang, Vietnam)	24

Foundations of Computer Science • Drug Repositioning by Bipartite Local Models Phuong H. Nguyen (Thuyloi University, Vietnam), Duc-Hau Le (Thuyloi University, Vietnam) • Hardware Trojan Threat and Its Countermeasures Xuan-Thuy Ngo (Secure-IC S.A.S, France), Van-Phuc Hoang (Le Quy Don Technical University, Vietnam), Han Le Duc (Le Quy Don Technical University, Vietnam)

• Linear time algorithm for maximum distance-\$k\$ matching in interval graphs

Viet Hung Tran (Hanoi University of Science and Technology, Vietnam), Nguyen Duc

Nghia (Hanoi University of Science and Technology, Vietnam, Vietnam), Phan-Thuan Do (Hanoi University of Science and Technology, Vietnam)

30

35

53

• User's perception on mental health applications: a qualitative analysis of user reviews

Kong Saoane Thach (Tra Vinh University, Vietnam)

47

• Adding External Features to Convolutional Neural Network for Aspect-based Sentiment Analysis

Nguyen Thi Xuan Huong (HaiPhong Private University, Vietnam), Vo Cong Hieu (Ton Duc Thang University, Vietnam), Le Anh Cuong (Ton Duc Thang University, Vietnam)

Modern Linguistics

• Response particles: The mysteries of yes and no	
Manfred Krifka (Leibniz-Zentrum Allgemeine Sprachwissenschaf, Germany)	60
• Polarity and the diachronic development of deontic modality in Chinese	
Barbara Meisterernst (National Tsing Hua University, Taiwan)	66
• Towards a Compositional Semantics for L+H*LH%	
Marie-Christine Meyer (ZAS, Germany)	7(

Computational Intelligence

• Deep Learning versus Traditional Classifiers on Vietnamese Students' Feedback Corpus

Phu Nguyen (University of Information Technology, VNU-HCM, Vietnam), Tham
Truong (University of Information Technology, VNU-HCM, Vietnam), Kiet Van
Nguyen (University of Information Technology, VNU-HCM, Vietnam), Ngan L.T.
Nguyen (University of Information Technology, VNU-HCM, Vietnam)
75

• A Coevolutionary approach for classification problems: Preliminary results

Van Truong Vu (Le Quy Don Technical University, Vietnam), Bui Lam (Le Quy Don University, Vietnam), Trung Nguyen (Liverpool John Moores University, United Kingdom (Great Britain))

81

• Two new concepts "Picture Fuzzy Rough Soft Sets" and "Picture Fuzzy Dynamic Systems" in Picture Fuzzy Systems

	of Science & Vietna	n National U1	niversity Han	oi, Vietnam)		87
	kt Generation from Vietnamese Paragraph			_	for Summar	rizing
	Trung Tran (Univers	ity of Informa	tion Technol	ogy, VNU - HCMO	· ·	ng 93
Computa	ational Intelligen	ce				
	ation Extraction in Vith Word-Level Atte		Гехt via Pie	ecewise Convolut	ion Neural Net	work
	Nhat Nguyen (University (University of Engin Vietnam)	of Engineerin	g and Techno	ology, VNU, Vietna	am), Hieu Vo	
• Dec	ep Learning for Aspe	ct Detection	on Vietnam	ese Reviews		
	Dang Van Thin (Uni Vietnam), Vu Duc N Vietnam), Kiet Van I Vietnam), Ngan L.T. Vietnam)	guyen (Unive Nguyen (Univ	ersity of Infor versity of Info	mation Technology ormation Technolog	y, VNU-HCM, gy, VNU-HCM,	, 104
	Hybrid Feature Sel Mellitus: KNHANES Huilin Zheng (Chung National University, Kwang Ho Park (Ch	2013-2015 gbuk National Korea), Ding	University, I kun Li (Chun	Korea), Hyun Woo gbuk National Uni	Park (Chungbuk versity, Korea),	
	University, Korea)					110
Commur	nications and Net	working				
• Rep	Deated Index Modula Le Huyen (Le Quy D University of Science Center, MobiFone C University, Vietnam)	Oon Technical e and Technol orporation, Vi	University, V logy, Vietnan	Vietnam), Vu-Duc I n), Minh-Tuan Le (MobiFone R&D	
	olementation of IDM and non-Gaussian No	•	ith Multiple	Access Channel		
	Bao Quoc Vuong (In (International Univer			UHCM, Vietnam),	Hung Ngoc Do	119
	formance Comparison BLE Beacons	on of Indoor	Positioning	Schemes Exploi	ting Wi-Fi AP	s and
	Younguk Yun (Kwar University, Korea), I (Hoseo Telnet, Korea	Dukhyun Ann	(Kwangwoon	n University, Korea	a), Sangsoo Kim	124

Bui Cuong (Institute of Mathematics, Hanoi, Vietnam), Pham Thong (VNU University

•	Some observations on Vietnamese demonstratives
	Phan Thi Huyen Trang (Vietnam National University, Vietnam), Dylan Tsai (National
	Tsing Hua University, Taiwan) 128
•	The Asymmetry of Topicalization: a View from Mandarin Chinese
	Wei-wen Roger Liao (Academia Sinica, Taiwan) 132
Softw	are Engineering
	Modelling Dynamic Information Flows: Extensions of LINQ with Norms
	Nguyen Hoang Thuan (Can Tho University of Technology, Vietnam), Tran Anh Tri
	(Can Tho University, Vietnam), David Swann (LINQ Ltd, Vietnam), Huu-Hoa Nguyen
	(Can The University Vietnam)
	(Call The Chrystally, Victually)
•	Electroencephalography Analysis Using Neural Network
	Ngan Vuong Thuy Nguyen (University of Science, VNU-HCM, Vietnam), Tuan Van
	Huynh (University of Science, VNU-HCM, Vietnam), Hanh Thi Hong Nguyen (VNU-
	University of Science Ho Chi Minh city, Vietnam) 144
•	The Influence of Icon Background Colors and Icon Symbols on Menu Item Selection for Smartphone
****	Lumpapun Punchoojit (Thammasat University, Thailand), Nuttanont Hongwarittorrn
	(Thammasat University, Thailand) 148
Comp	outational Intelligence
•	An Efficient Parallel Algorithm for Mining Both Frequent Closed and Generator
	Sequences on Multi-core Processors
	Hai Duong (University of Dalat, Vietnam), Tin Truong (University of Dalat, Vietnam),
	Bac Le (University of Science. VNU HCM, Vietnam) 154
	Building mathematical models applied to UTXOs selection for objective transactions
	Nguyen Huy (University of Technology, Vietnam), Hong-Son Trang (Hoa Sen
	University, Vietnam), Thinh Nguyen (Ho Chi Minh City University of Technology,
	Vietnam), Nguyen Huynh Tuong (Faculty of Computer Science & Engineering, Ho Chi
	Minh city University of Technology, Vietnam), Thanh Van Le (Ho Chi Minh City
	University of Technology, Vietnam) 160
•	An Ensemble of Shallow and Deep Learning Algorithms for Vietnamese Sentiment Analysis
	Nguyen Quan (Le Quy Don Technical University, Vietnam), Uy Nguyen (Le Quy Don
	Technical University, Vietnam) 165
•	On Temporal Cluster Analysis for Early Identifying In-trouble Students in an Academic Credit System
	Chau Le (Ho Chi Minh City University of Technology, Vietnam), Chau Thi Ngoc Vo
	Minh City University of Technology, Vietnam) 171
	Chau Le (Ho Chi Minh City University of Technology, Vietnam), Chau Thi Ngoc Vo (Ho Chi Minh City University of Technology, Vietnam), Phung Hua Nguyen (Ho Chi Minh City University of Technology, Vietnam) 171

Communications and Networking

• A role-based statistical mechanism for DDoS attack detection in SDN

Phom (University of Information Technology, Vietnam)	m), Van-Hau
Pham (University of Information Technology, Vietnam)	177
• Hardware Implementation of a MIMO Channel Emulator for high 802.11ac	n speed WLAN
Van Tien Tran (University of Information Technology, Vietnam), Man	h Tien Tran
(University of Information Technology, Vietnam), Duc Khai Lam (University of Information Technology, Vietnam)	iversity of
Information Technology, VNU-HCM, Vietnam)	183
• Enhanced Spread Spectrum OFDM-IM with Rotated Zadoff-Chu Seq	uences
Thien Van Luong (Queen's University Belfast, United Kingdom (Great	* * * * * * * * * * * * * * * * * * * *
Youngwook Ko (Queen's University Belfast, United Kingdom (Great I	***
Ngo (Hanoi University of Science and Technology, Vietnam), Minh-T	
(MobiFone R&D Center, MobiFone Corporation, Vietnam), Le Huyen	` ` •
Technical University, Vietnam), Nam Xuan Tran (Le Quy Don Techni	•
Vietnam)	189
• Nonnegative Tensor Decomposition for EEG Epileptic Spike Detection	on
Nguyen Thi Anh Dao (University of Technology and Logistics, Vietna	m), Le Trung
Thanh (Vietnam National University, Hanoi, Vietnam), Nguyen Linh-	Trung (Vietnam
National University, Hanoi, Vietnam), Vu Ha Le (UET, Vietnam)	194
puter Vision and Intelligent Systems	
Occluded Image Recognition with Extended Nonnegative Matrix Fact	
Viet-Hang Duong (BacLieu University, Vietnam), Manh-Quan Bui (N	
University, Taiwan), Jia-Ching Wang (National Central University, Ta	iwan) 200
• Region-based deformation transfer	
Phong Khac Do (Vietnam National University, Hanoi, Vietnam), Chau	Ma Thi (VNU
University of Engineering and Technology, Vietnam), Giang Cao (Can	Tho University,
	205
Vietnam), An Thi Thu Nguyen (Can Tho University, Vietnam)	
Vietnam), An Thi Thu Nguyen (Can Tho University, Vietnam) • Planar Object Recognition For Bin Picking Application	
	University of
Planar Object Recognition For Bin Picking Application	-
 Planar Object Recognition For Bin Picking Application Đức Hạnh Lê (Supervisor, Vietnam), Le Minh Duc (Ho Chi Minh City Technology, Vietnam) A Frame-work assisting the Visually Impaired People: Common C	211
 Planar Object Recognition For Bin Picking Application Đức Hạnh Lê (Supervisor, Vietnam), Le Minh Duc (Ho Chi Minh City Technology, Vietnam) A Frame-work assisting the Visually Impaired People: Common Cand Pose Estimation in Surrounding Environment 	211 Object Detection
 Planar Object Recognition For Bin Picking Application Đức Hạnh Lê (Supervisor, Vietnam), Le Minh Duc (Ho Chi Minh City Technology, Vietnam) A Frame-work assisting the Visually Impaired People: Common C and Pose Estimation in Surrounding Environment Le Hung (International Research Institute MICA - HaNoi University S 	211 Dbject Detection cience and
 Planar Object Recognition For Bin Picking Application Đức Hạnh Lê (Supervisor, Vietnam), Le Minh Duc (Ho Chi Minh City Technology, Vietnam) A Frame-work assisting the Visually Impaired People: Common Cand Pose Estimation in Surrounding Environment 	211 Dbject Detection cience and Hanoi
 Planar Object Recognition For Bin Picking Application Đức Hạnh Lê (Supervisor, Vietnam), Le Minh Duc (Ho Chi Minh City Technology, Vietnam) A Frame-work assisting the Visually Impaired People: Common Cand Pose Estimation in Surrounding Environment Le Hung (International Research Institute MICA - HaNoi University S Technology, Vietnam), Hai Vu (International Research Institute MICA) 	211 Dbject Detection cience and Hanoi etnam National
 Planar Object Recognition For Bin Picking Application Đức Hạnh Lê (Supervisor, Vietnam), Le Minh Duc (Ho Chi Minh City Technology, Vietnam) A Frame-work assisting the Visually Impaired People: Common C and Pose Estimation in Surrounding Environment Le Hung (International Research Institute MICA - HaNoi University S Technology, Vietnam), Hai Vu (International Research Institute MICA University of Science and Technology, Vietnam), Nguyen T. Thuy (Vietnam) 	211 Dbject Detection cience and Hanoi etnam National
 Planar Object Recognition For Bin Picking Application Đức Hạnh Lê (Supervisor, Vietnam), Le Minh Duc (Ho Chi Minh City Technology, Vietnam) A Frame-work assisting the Visually Impaired People: Common Cand Pose Estimation in Surrounding Environment Le Hung (International Research Institute MICA - HaNoi University S Technology, Vietnam), Hai Vu (International Research Institute MICA University of Science and Technology, Vietnam), Nguyen T. Thuy (Vi University of Agriculture, Vietnam) 	Dbject Detection cience and Hanoi etnam National 216
 Planar Object Recognition For Bin Picking Application Đức Hạnh Lê (Supervisor, Vietnam), Le Minh Duc (Ho Chi Minh City Technology, Vietnam) A Frame-work assisting the Visually Impaired People: Common C and Pose Estimation in Surrounding Environment Le Hung (International Research Institute MICA - HaNoi University S Technology, Vietnam), Hai Vu (International Research Institute MICA University of Science and Technology, Vietnam), Nguyen T. Thuy (Vi University of Agriculture, Vietnam) Bilingualism and the Lifespan: Young Adult Heritage Speakers of Spanning Agriculture and Spanning Adult Heritage Speakers of Spanning Adult Heritage Spanning Adult Heritage Speakers of Spanning Adult Heritage Spanning A	211 Dbject Detection cience and Hanoi etnam National 216
 Planar Object Recognition For Bin Picking Application Đức Hạnh Lê (Supervisor, Vietnam), Le Minh Duc (Ho Chi Minh City Technology, Vietnam) A Frame-work assisting the Visually Impaired People: Common Cand Pose Estimation in Surrounding Environment Le Hung (International Research Institute MICA - HaNoi University S Technology, Vietnam), Hai Vu (International Research Institute MICA University of Science and Technology, Vietnam), Nguyen T. Thuy (Vi University of Agriculture, Vietnam) 	211 Dbject Detection cience and Hanoi etnam National 216

Phan The Duy (University of Information Technology, VNU-HCM, Vietnam), Do Thi

	ssessing the Readability of Literary Texts in Vietnamese Textbooks	
	An-Vinh Luong (HoChiMinh city University of Science, Vietnam), Diep Nguyer	ı
	(SaiGon Technology University, Vietnam), Dinh Dien (University of Natural Sci	ences,
	Vietnam)	231
nput	tational Intelligence	
	n Efficient Hardware Implementation of Artificial Neural Network bas Stochastic Computing	sed on
	Duy-Anh Nguyen (VNU University of Engineering and Technology, Vietnam), I	Huy-
	Hung Ho (VNU University of Engineering and Technology, Vietnam), Duy-Hieu	•
	(VNU University of Engineering and Technology (VNU-UET), Vietnam), Xuan-	
	Tran (VNU University of Technology and Engineering, Vietnam)	237
• In	tegrating Grammatical Features into CNN Model for Emotion Classification	
	Thi-Thanh-Thuy Huynh (Ton Duc Thang University, Vietnam), Le Anh Cuong (Ton
	Duc Thang University, Vietnam)	243
• A1	n Effective Similarity Measure for Neighborhood-based Collaborative Filterin	
	Tan Nghia Duong (Hanoi University of Science and Technology, Vietnam), Viet	
	Than (Hanoi University of Science and Technology, Vietnam), Trong Hiep Tran	*
	University of Science and Technology, Vietnam), Quang Hieu Dang (Hanoi University of Science and Technology, Vietnam), Quang Hieu Dang (Hanoi University of Science and Technology, Vietnam), Quang Hieu Dang (Hanoi University of Science and Technology), Vietnam), Quang Hieu Dang (Hanoi University of Science and Technology), Vietnam), Quang Hieu Dang (Hanoi University of Science and Technology), Vietnam), Quang Hieu Dang (Hanoi University of Science and Technology), Vietnam), Quang Hieu Dang (Hanoi University of Science and Technology), Vietnam), Quang Hieu Dang (Hanoi University of Science and Technology), Vietnam), Quang Hieu Dang (Hanoi University of Science and Technology), Vietnam), Quang Hieu Dang (Hanoi University of Science and Technology), Vietnam (Hanoi University of Science and Technol	-
	of Science and Technology, Vietnam), Nguyen D. Minh (HUST, Vietnam), Phan	n Manh
	Hung (VNPT Technology, Vietnam)	250
-	peech perception based on mapping speech to image by using convolution network	neural
	Nguyen Quang Trung (Human Machine Interaction Laboratory, University of	
	E ' ' OE 1 1 INTILITAL'IL' \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \	
	Engineering & Technology, VNU Ha Noi, Vietnam), The Duy Bui (Human Mac	hine
	Interaction Laboratory, Vietnam)	hine 255
• Co	Interaction Laboratory, Vietnam)	
•Co	Interaction Laboratory, Vietnam) ollecting Chinese-Vietnamese Texts From Bilingual Websites	255
• Co	Interaction Laboratory, Vietnam) ollecting Chinese-Vietnamese Texts From Bilingual Websites Minh Trinh (Ton Duc Thang University, Vietnam), Phuoc Tran (Faculty of IT, T	255 DTU
	Interaction Laboratory, Vietnam) ollecting Chinese-Vietnamese Texts From Bilingual Websites Minh Trinh (Ton Duc Thang University, Vietnam), Phuoc Tran (Faculty of IT, T & NLP-KD, Vietnam), Nhung Tran (Ton Duc Thang University, Vietnam)	255
	Interaction Laboratory, Vietnam) ollecting Chinese-Vietnamese Texts From Bilingual Websites Minh Trinh (Ton Duc Thang University, Vietnam), Phuoc Tran (Faculty of IT, T	255 DTU
ı mu •Er	Interaction Laboratory, Vietnam) ollecting Chinese-Vietnamese Texts From Bilingual Websites Minh Trinh (Ton Duc Thang University, Vietnam), Phuoc Tran (Faculty of IT, T & NLP-KD, Vietnam), Nhung Tran (Ton Duc Thang University, Vietnam)	255 DTU 260
ı mu •Er	Interaction Laboratory, Vietnam) ollecting Chinese-Vietnamese Texts From Bilingual Websites Minh Trinh (Ton Duc Thang University, Vietnam), Phuoc Tran (Faculty of IT, T & NLP-KD, Vietnam), Nhung Tran (Ton Duc Thang University, Vietnam) unications and Networking nergy-Efficient and Low Complexity Channel Coding for Wireless Body Networks Hieu T. Nguyen (University in Southeast Norway, Norway), Thuy V. Nguyen (Personal Coding Cod	DTU 260 y Area osts
mu •Er	Interaction Laboratory, Vietnam) ollecting Chinese-Vietnamese Texts From Bilingual Websites Minh Trinh (Ton Duc Thang University, Vietnam), Phuoc Tran (Faculty of IT, T & NLP-KD, Vietnam), Nhung Tran (Ton Duc Thang University, Vietnam) inications and Networking nergy-Efficient and Low Complexity Channel Coding for Wireless Body Networks	DTU 260
ımu •Er	Interaction Laboratory, Vietnam) ollecting Chinese-Vietnamese Texts From Bilingual Websites Minh Trinh (Ton Duc Thang University, Vietnam), Phuoc Tran (Faculty of IT, T & NLP-KD, Vietnam), Nhung Tran (Ton Duc Thang University, Vietnam) unications and Networking nergy-Efficient and Low Complexity Channel Coding for Wireless Body Networks Hieu T. Nguyen (University in Southeast Norway, Norway), Thuy V. Nguyen (Personal Coding Cod	DTU 260 y Area osts 265
ımu •Er	Interaction Laboratory, Vietnam) ollecting Chinese-Vietnamese Texts From Bilingual Websites Minh Trinh (Ton Duc Thang University, Vietnam), Phuoc Tran (Faculty of IT, T & NLP-KD, Vietnam), Nhung Tran (Ton Duc Thang University, Vietnam) inications and Networking nergy-Efficient and Low Complexity Channel Coding for Wireless Body Networks Hieu T. Nguyen (University in Southeast Norway, Norway), Thuy V. Nguyen (Pand Telecommunications Institute of Technology, Vietnam)	DTU 260 y Area osts 265 ADCs
ımu •Er	Interaction Laboratory, Vietnam) ollecting Chinese-Vietnamese Texts From Bilingual Websites Minh Trinh (Ton Duc Thang University, Vietnam), Phuoc Tran (Faculty of IT, T & NLP-KD, Vietnam), Nhung Tran (Ton Duc Thang University, Vietnam) unications and Networking nergy-Efficient and Low Complexity Channel Coding for Wireless Body Networks Hieu T. Nguyen (University in Southeast Norway, Norway), Thuy V. Nguyen (Pand Telecommunications Institute of Technology, Vietnam) ully Digital Background Calibration Technique for Channel Mismatches in TI	255 DTU 260 y Area osts 265 ADCs e Quy

Dung Hoang (University of Education, Ho Chi Minh City, Vietnam), Mai Le Nguyen

227

Hoang (University of Education, Vietnam)

Computer Vision and Intelligent Systems

 Machine Learning Based-distributed Optimal Control Algorithm for Multiple Nonlinear Agents with Input Constraints
Luy Tan Nguyen (Industrial University of Ho Chi Minh City, Vietnam), Dang Nguyen (Industrial University of Ho Chi Minh City, Vietnam), Minh Dang (Industrial University of Ho Chi Minh City, Vietnam), Vinh Tran (Industrial University of Ho Chi Minh City, Vietnam) 276
• An Intelligent Support System for the knowledge evaluation in high-school mathematics by multiple choices testing
Thanh Mai (Ho Chi Minh city Open University, Vietnam), Hien Nguyen (University of Information Technology, Vietnam), Trung Le (Vinh Long College of Economics and Finance, Vinh Long, Vietnam), Vuong Pham (University of Information Technology, VNU-HCM, Vietnam) 282
• A new ensemble approach for hyper-spectral image segmentation
Le Cam Binh (Academy of Military Science and Technology, Vietnam), Pham Van Nha (Academy of Military Science and Technology & Le Quy Don University, Vietnam), Long Thanh Ngo (Le Quy Don University, Vietnam), Pham The Long (Le Quy Don University, Vietnam)
Preliminary Result of 3D City Modelling For Hanoi, Vietnam
Anh Phan (VNU University of Engineering and Technology, Vietnam), Vu Chu (Center of Multidisciplinary Integrated Technology for Field Monitoring, Vietnam), Hung Bui (VNU University of Engineering and Technology, Vietnam), Thanh Nguyen (VNU University of Engineering and Technology, Vietnam), Nguyen Viet Ha (VNU Ha Noi, Vietnam)
• Vision-based Inspection System for Leather Surface Defect Detection and Classification
Hoang-Quan Bong (Kiengiang Vocational College, Vietnam), Truong Quoc Bao (CanTho University, Vietnam), Huu-Cuong Nguyen (Can Tho University, Vietnam), Minh Triet Nguyen (CanTho University, Vietnam)
Modern Linguistics
Linguistic barriers to syllogistic reasoning
Andreas Haida (The Hebrew University of Jerusalem, Israel), Luka Crnič (The Hebrew University of Jerusalem, Israel), Yosef Grodzinsky (The Hebrew University of Jerusalem, Israel) 305
Parsing out in English and Vietnamese
Nigel Duffield (Konan University, Japan) 311
• The Participant-Pronoun Restriction: English and Vietnamese Tue Trinh (University of Wisconsin-Milwaukee, USA), Hubert A Truckenbrodt

Poster Session

• Improving Phonetic Recognition with Sequence-length Standardized MFCC Features and Deep Bi-directional LSTM

317

(Leibniz-Zentrum Allgemeine Sprachwissenschaft, Germany)

Pham Van Toan (Framgia Inc, Vietnam), Nguyen Thanh Hau (Framgia Inc, Vietna	m),
Minh Thanh Ta (Le Quy Don University, Vietnam)	322
• Improving the 3D model classification based on selecting proper features	
Nong Thi Hoa (Thai Nguyen University of ICT, Vietnam), Nguyen Van Tao	
(University of Information & Communication Technology, Vietnam), Dinh Thi Th	anh
Uyen (Thai Nguyen University of Agriculture and Forestry, Vietnam)	326
• Towards An Educational Music Processor for Folk and Popular Musics	
Anh-Thu G. Phan (Columbia University & Temple University, USA), Nhan T Ngo	+
(New York University & Temple University, Center for Vietnamese Philosophy,	220
Culture and Society, USA)	330
• Study on Cloud computing and Emergence of Internet of the Thing in Industry	
Symphorien Yoki Donzia (South Korea, Korea), Haeng Kon Kim (Daegu Catholic	
University, Korea), Bo Yeon Shin (Seoul, Korea)	334
•FVI: An End-to-end Vietnamese Identification Card Detection and Recognition	on in
Images	<i>J</i> 11 111
Liem Hoang (FPT Technology Research Institute, FPT University, Vietnam),	
Hoang Vu Dang (FPT Technology Research Institute, Vietnam)	338
Troung vir Bung (111 Teenmoregy resource matrices, vietnam)	
• A Practical Solution to the ACM RecSys Challenge 2018	
Tan Nghia Duong (Hanoi University of Science and Technology, Vietnam), Viet D	
Than (Hanoi University of Science and Technology, Vietnam), Trong Hiep Tran (H	Ianoi
University of Science and Technology, Vietnam), Thi Hong Anh Pham (Hanoi	
University of Science and Technology, Vietnam), Van Hoang Anh Nguyen (Hanoi	
University of Science and Technology, Vietnam), Hoang Nam Tran (Thanglong Hi	igh
School, Hanoi, Vietnam)	341
• A Reconfigurable Multi-function DMA Controller for High-Performance Comp	uting
Systems	
Hung K. Nguyen (VNU University of Engineering and Technology, Vietnam), Kho	oi
Dong (VNU University of Engineering and Technology, Vietnam), Xuan-Tu Tran	
(VNU University of Technology and Engineering, Vietnam)	344
A Door I coming Model for Extracting User Attributes from Convergetional Tax	t a
• A Deep Learning Model for Extracting User Attributes from Conversational Text	
Pham Quang Nhat Minh (Alt Vietnam Co., Ltd., Vietnam), Tuan-Anh Nguyen (Alt Vietnam)	
Vietnam Co., Ltd., Vietnam), Tuan Duc Nguyen (Alt Vietnam Co., Ltd., Vietnam)	350
Building Vietnamese Linguistic Resources for Social Network Text Analysis	
Tuyen Nguyen (FPT Corporation, Vietnam), Luong Vu (Vietnam Lexicography Co	enter,
Vietnam), Phuong Le-Hong (Vietnam National University, Hanoi & FPT Technology	ogy
Research Institute, FPT Corporation, Vietnam)	354
• Lead Engagement by Automated Real Estate Chatbot	
Quan Thanh Tho (HCMUT, Vietnam), Trung Trinh (Ho Chi Minh City University	of
Technology, Vietnam), Dang Ngo (Ho Chi Minh City University of Technology,	01
Vietnam), Hon Pham (Ho Chi Minh City University of Technology, Vietnam), Lor	າσ
Hoang (Atomic Vietnam Co., LTD, Vietnam), Hung Hoang (Atomic Vietnam Co.,	_
I TD Vietnam) Thanh Thai (Ho Chi Minh City University of Technology Vietnam	

Computational Intelligence

• A	Comparative	Study of 1	Neural	Network	Models	for S	Sentence	Classifica	tion
-----	-------------	------------	--------	---------	--------	-------	----------	------------	------

Phuong Le-Hong (Vietnam National University, Hanoi & FPT Technology Research Institute, FPT Corporation, Vietnam), Le Anh Cuong (Ton Duc Thang University, Vietnam)

360

• A Hybrid Approach to Paraphrase Detection

Phuc H. Duong (Ton Duc Thang University, Vietnam), Nguyen Thanh Hien (TDT, Vietnam), Hieu Duong (Ho Chi Minh City University of Technology, Vietnam), Khoa Ngo (NewAI Research, Vietnam), Dat Ngo (NewAI Research, Vietnam)

• A review of feature indexing methods for fast approximate nearest neighbor search
The-Anh Pham (Hong Duc University, Vietnam), Van-Hao Le (Hong Duc University
(HDU) Thanh Hoa, Vietnam), Dinh-Nghiep Le (Hong Duc University (HDU), Thanh
Hoa, Vietnam)

372

• Building a Spelling Checker for Documents in Khmer Language

Trần Văn Nam (Trường Đại học Trà Vinh, Vietnam), Nguyen Thi Hue (Tra Vinh University, Vietnam), Huy Khanh Phan (Danang College of Technology, Vietnam) 378

Computer Vision and Intelligent Systems

• A solution based on combination of RFID tags and facial recognition for monitoring systems

Hoang Van Dung (Quang Binh University, Vietnam), Van-Dat Dang (Quang Binh University, Vietnam), Tien-Thanh Nguyen (Science and Technology Department of Quang Binh, Vietnam), Diem-Phuc Tran (Duy Tan University, Vietnam)

384

• An effective implementation of Gaussian of Gaussian descriptor for person reidentification

Thuy-Binh Nguyen (Hanoi University of Science and Technology, Vietnam), Duc-Long Tran (MICA, HUST, Vietnam), Thi-Lan Le (MICA, HUST, Vietnam), Pham Thanh Thuy (MICA Institute (HUST - CNRS/UMI 2954 - INP Grenoble), Vietnam), Huong-Giang Doan (MICA, HUST, Vietnam)

• Hybrid discriminative models for banknote recognition and anti-counterfeit

Hoang Van Dung (Quang Binh University, Vietnam), Hoang-Thanh Vo (Quang Binh University, Vietnam) 394

• Total Variation L1 Fidelity Salt-and-Pepper Denoising with Adaptive Regularization Parameter

Dang N. H. Thanh (Hue College of Industry, Vietnam), V. B. Surya Prasath (Cincinnati Children's Hospital Medical Center & University of Cincinnati, USA), Le Thi Thanh (Ho Chi Minh city University of Transport, Vietnam) 400

Computational Intelligence

• Semi-supervised method with Spatial weights based Possibilistic fuzzy C-means clustering for Land-cover Classification

Dinh-Sinh Mai (Le Quy Don Technical University, Vietnam), Thanh Long Ngo (Le Quy Don University, Vietnam) 406

• Vietnamese Keyword Extraction Using Hybrid Deep Learning Methods

Hung Thanh Bui (Thu Dau Mot University & Data Analytics & Artificial Intelligence Laboratory, Vietnam) 412

• Empirical Evaluation of Link Prediction Methods in Social Networks

Ba-Hien Tran (John von Neumann Institute, Vietnam National University, Ho Chi Minh City, Vietnam) 418

Computer Vision and Intelligent Systems

• A Predictive Model for ECG Signals Collected from Specialized IoT Devices using Deep Learning

Duy Tran (Ho Chi Minh City University of Technology, Vietnam), Thanh Vo (Ho Chi Minh City University of Technology, Vietnam), Dung Nguyen (Ho Chi Minh City University of Technology, Vietnam), Quan Nguyen (Ho Chi Minh City University of Science, Vietnam), Liem Huynh (GTOPIA Joint Stock Company, Vietnam), Ly Le (International University, USA), Hai Do (Ho Chi Minh City International University, Vietnam), Tho Quan (Bach Khoa University, Vietnam), Trung Mai (Bach Khoa University, Vietnam)

• Joint Image Deblurring and Binarization for License Plate Images using Deep Generative Adversarial Networks

Van-Giang Nguyen (Le Quy Don Technical University, Vietnam), Duy Long Nguyen (Le Quy Don Technical University, Vietnam) 430

 Deep Learning-based Multiple Objects Detection and Tracking System for Socially Aware Mobile Robot Navigation

Do Nam Thang (Le Quy Don Technical University, Vietnam), Nguyen Lan Anh (Le Quy Don Technical University, Vietnam), Trung Dung Pham (Military Technical Academy, Vietnam), Truong Dang Khoa (Le Quy Don Technical University, Vietnam), Nguyen Huu Son (Le Quy Don Technical University, Vietnam), Pham Van Nguyen (Le Quy Don Technical University, Vietnam), Nguyen Hiep (Le Quy Don University, Vietnam), Vu Duc Truong (Le Quy Don Technical University, Vietnam), Dinh Hong Toan (Le Quy Don Technical University, Vietnam), Nguyen Manh Hung (Le Quy Don Technical University, Vietnam), Trung Dung Ngo (University of Prince Edward Island, Canada), Xuan-Tung Truong (Le Quy Don Technical University, Vietnam)

Towards an Educational Music Processor for Folk and Popular Musics

Anh-Thu G. Phan, M.A., Ed.M. Music Education

Teachers College

Columbia University

New York, NY USA

agp2132@tc.columbia.edu

Center for Vietnamese Philosophy, Culture & Society
Temple University
Philadelphia, PA USA
anh.thu.phan@temple.edu

Thanh-Nhan Ngo, Ph.D. Linguistics

Courant Computer Science, Medical Language Processing

New York University

New York, NY USA

nhan@cs.nyu.edu

Center for Vietnamese Philosophy, Culture & Society
Temple University
Philadelphia, PA USA
nhan@temple.edu

Abstract—This paper describes an educational musical processor that takes spectrographic data of a sonic object and turns them into a series of meaningful layers associated with different musical knowledge representations, in such a way that it can be understood, reproduced, played, compared, and taught by everyone across cultures, regardless of their musical backgrounds. Any music audio file can be used as input. Within the scope of this paper, the authors focus on processing of musical audio files onto common graphic platform of physical sound properties, in Hertz, Decibels, and milliseconds, so that culturally dependent musical units such as notes, beats, measures, phrases, chords, and sections can be viewed in separate layers. Syntactic techniques, such as frequency of occurrences, and adjacency are applied to musical units, such as pitches and musical chords. They are key pitches in context and key chords in context. The results are then mapped onto circles of fifths which reveal distinct patterns of each song, each section of one song, of each artist, each genre, and each culture. Semiautomatic generation of layers of annotations on top of the spectrogram helps teachers to quickly discover and/or compare distinctive features of a song, while preparing lessons. Learners of all levels can choose the most prominent patterns of the song to learn. This can also advance methods for preservation for further studies of sonic objects in the future.

Keywords—music processor, music representation, audio files, frequency, pitch, melody, chord, harmonic progression, musical form, circle of fifths, frequency of occurrence, distinctive feature, key words in context, key pitches in context, key chords in context.

I. INTRODUCTION

For years, social sciences and humanities rarely involve the study of music. With the advent of artificial intelligence, especially an extensive understanding of natural language processing, and the massive use of digitized sonic objects, it is possible to imagine a platform to present, preserve, and analyze musical and linguistic elements using string grammar, as in [1]. Music and songs can be thought of as linear strings of pitches or syllables from performers to listeners. Music and linguistic units, such as pitches and syllables, are inherently not well-tempered. Notating human music in western notation has stripped away many essential music features, as in [2].

Adding these once-lost features back to their context (adjacent pitches and syllables) reveals culturally dependent characteristics of sonic artefacts, as in [3]. In this paper, natural and well-tempered pitches are presented in different layers or superimposed on each other to help further analysis and comparison, as in [4] and [5]. Linearity, adjacency, regularity, structure, and unit-oriented properties of music from pitches, phrases, sections are similar enough, to that of natural language processing (NLP), as in [6]. This NLP processor helps sketch major components of this educational music processor (EMP).

This paper is largely divided into two parts, one part using technologies to convert musical objects into physical data, measured by Hertz (Hz), Decibels (dBs), and milliseconds (ms). From these units, a layer of musical clefs and layers of interpretive manual insertion of beats, measures, phrases, sections, lyrics, and chords, are overlaid laid. The second part is dedicated to describing an EMP via a test case. It displays frequencies of occurrences of pitches and chords, and strings of adjacent pitches and adjacent key chords. The results expose the unprecedented characteristics of the test song.

II. OBTAINING DATA

Data of this paper can be divided into two categories: phenomenal and conventional (or knowledge). Phenomenal data are physical signals such as time, frequency, and intensity. Conventional data are interpretive symbols and knowledge that have been used in music and languages in different cultures. Examples of conventional data in this paper are musical beats, pitches, phrases, sections, quarter notes, measures, treble and bass clefs, and chords, which exist only in culturally learned contexts.

III. TECHNOLOGIES USED

Top-of-the-art technologies are tested and chosen based on three criteria: user-friendliness, cost, and standardization.

A. Sound Analysis Software

- 1) Sonic Visualiser, is a free software [7] that allows visualized representations of audio input such as soundwaves and spectrograms. It allows manually or plug-in overlaid annotations on top of one another. For examples: Chordino Chord Estimation, generates an SVL output¹ containing frames (defined in IV.A. below) associated with the chord labels [8]. Melodia Melody Extraction generates an SVL output containing frames associated with the melodic pitches [9]. Queen Mary Note Onset Detector/Percussive Onset Detector generates an SVL output containing frames associated with an onset of a percussive sound [10]. BBC Intensity generates an SVL output containing frames associated with an intensity value [11].
- 2) MuseScore (offline) is a free standalone open-source software for music notation [12]. The companion MuseScore (online), is an online platform allowing users to share and edit sheet music, and connect to other websites, or through YouTube videos [13]. MuseScore offline and MuseScore online versions are used in tandem by the EMP to create a panel of western musical notation.
- 3) AnthemScore is a standalone which suggests a transcription of an audio file (MP3, WAV, etc.) into sheet music using western notations [14]. It suggests pitches based dominant frequencies' amplitudes of the song's spectrogram.

B. The educational musical processor (EMP)

- 1) The EMP is coded in PHP 7.2 with SVG 1.1², follows the international, world wide web, archival, librarian and multilingual standards, Unicode and ISO/IEC 10646 [15], world wide web standards are HTML [16], XML [17], URI and MusicXML.³ The EMP also adheres to the Open Archive Initiative/Object Reuse and Exchange (OAI/ORE) [18] with Dublin Core Metadata Initiative (DCMI) [19].
- 2) The EMP runs about 6,000 lines of codes on a Dell T1650 workstation, with Ubuntu 18.04.1.

IV. A TEST CASE: DESPACITO BY LUIS FONSI 2017

The video version of *Despacito* by Luis Fonsi ft. Daddy Yankee (hereafter, simply *Despacito*)⁴ was chosen to guide the design of our folk and popular music processor. The song is the most viewed YouTube video.

A. Frame

Frame is a common feature that pairs all data collected. Each frame is 1/44,100 second or 10/441 millisecond. 44,100 Hz is one of the standard digital sample rates.

B. Spectrographic data plotted as the background layer

Despacito is 4 minutes and 42 seconds long. AnthemScore reads Despacito and generates a spreadsheet of spectrographic data. Data can either be obtained from every 1 ms to every 10

ms, in time, and from every 5 cents to every 25 cents, in pitch. The authors chose to test two sets of data: (a) a fine set of 1 ms by 25 cents in a spreadsheet of 351 pitch columns by 281,522 one-ms rows; and (b) a crude set of 10 ms by 25 cents in a spreadsheet of 351 pitch columns by 28,153 ten-ms rows. Each cell contains an intensity value in dBs. The former spreadsheet is 774 MB and the latter, 94 MB, in size.

The EMP reading of the fine spreadsheet took 130 ms, with a pitch range from 26.7171 to 4,308.66 Hz, and intensity range from 0 to 17,872 dBs. The crude spreadsheet took 15.5 ms to read, with the intensity range from 0 to 4,246.1 dBs.

C. Musical grand staff and piano keyboard as SVG Layer 1

The 351 pitch values are set on the vertical axis of *Despacito* spectrographic graph, *Layer 0* (Fig. 1), ranging from C^0 (0 *cent* at 26.7171 Hz) to B^8 (10,700 *cents* at 4,366.08 Hz). The pixel colors, from deep blue, lowest, to white, strongest, represent the intensity in dBs of the song. This allows a treble and a bass clef, *Layer 1*, with a piano 88 keys from A^0 (900 *cents*) to C^8 (9,600 *cents*) to be drawn on top of *Layer 0*. The staff lines run along the horizontal time axis.

D. Layer 2: Labelling of Beats, Measures, Chords, Phrases and Sections on Despacito spectrogram

These data are generated by *Sonic Visualiser* in frames with manual inputs, in SVL format. They are superimposed on $Layer\ 0$ and $Layer\ 1$ to show how human perception paired with the physical data.

On top of each panel of spectrogram, a bar graph of dB values at each frame in the song is drawn in red with white dots underfoot representing the percussive onsets (Fig. 2).

The beats are shown in yellow arrows inside red transparent measure boxes and the chords, in cyan, at their onset time positions (Fig. 3).

At the bottom of *Layer 0*, the phrase boxes in orange contain their lyrics in black, while the lowest light brown bars represent sections in the song (Fig. 4).

Manual markings of *Despacito* result in 83 phrases, 113 lyric expressions, 22 chords, and 26 sections (representing 11 music forms labelled from A to L).

E. Layers for collections of pitch intensity of each beat, measure, chord, phrase and section.

Specifically, *Layer 3* shows dBs by measures, *Layer 4*, dBs by beats, *Layer 5*, dBs by chords, *Layer 6*, dBs by phrases and *Layer 7*, dBs by sections. Fig. 5 (left) shows *Layer 5*.

F. Layer 8, music sheet of Despacito according to AnthemScore

This is an estimate by AnthemScore. The notes are displayed over Layer 0, Layer 1 and Layer 2 to pair the estimates with the spectrographic data and other manual data.

The sums of all pitch dB values within a beat, a measure, a chord, a phrase and a section, reveal the strongest pitches within, evidence of played notes, marked on the vertical axis.

¹ SVL is an XML Sonic Visualiser data exchange format.

² At https://www.w3.org/TR/2011/REC-SVG11-20110816/.

³ MusicXML 3.0, in XML format for exchanging digital sheet music, see https://www.w3.org/2017/12/musicxml31/.

⁴ At https://www.youtube.com/watch?v=kJQP7kiw5Fk.

These layers are intended to reveal also differences in the manual perception against the raw spectrographic data.

G. Layer 9, AnthemScore estimation is displayed through MuseScore applications.

This layer is the most interactive, by *AnthemScore*. Viewers can play portions of *Despacito*, section by section. They can also play back the sheet music, section by section.

Layer 0 to 9, takes about 235 seconds to produce by the EMP, viewed at http://mlp.cs.nyu.edu/vietmusic/au2spec.php.

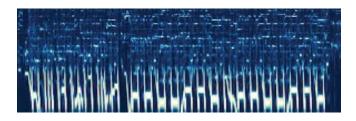


Fig. 1. A segment of spectrogram of Despacito's recording.



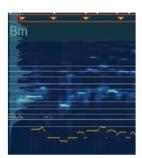
Fig. 2. An example of *Despacito* intensity and percussive onsets.



Fig. 3. An example of *Despacito* markings: down arrows for beats, rectangular boxes for measure, cyan letters for chord symbols.



Fig. 4. An example of *Despacito* markings: orange rectangular boxes for lyrics and light brown rectangular boxes for sections.



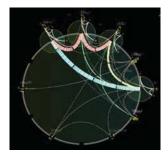


Fig. 5. Left, an example of Layer 5: sum of dBs by Chords B minor over *Despacito* spectrogram image. Right, the chord procession kcic(2) with their frequencies of occurrences in thickness of over the circle of fifths.



Fig. 6. An example of music notation on top of the spectrogram

I. Pitch and chord frequencies of occurrences, and key chords in context kcic(n), and their visual representations on the Circles of Fifths.

AnthemScore estimates that Despacito has a range of 47 pitches, from B⁰ to C#⁷: D⁴ occurs 508 times, B³, 353 times, F#⁴, 241 times, A³, 201 times, E⁴, 173 times, and so on. They are drawn on the music clefs, as well as on a circle of fifths with their frequencies as radii. This shows that even though the song has all 12 pitch classes, they strongly lean on 4 central pitches (highest frequencies of occurrence), while others are just decorative. Further analysis confirms this observation.

If the song is estimated section by section, it turns out that *AnthemScore* now produces a range of 60 pitches, from B⁰ to D⁷. The same 4 notes are still central to the song: D⁴, occurs 612 times, B³, 433 times, F#⁴, 292 times, and A³, 243 times, and so on. Their class positions are more prominent in the circle of fifths. The same analysis on frequencies of occurrences of pitches done between sections shows the tendencies of section movements towards each other, serving as clues to identification of musical forms.

Here the concept of key pitches in context kpic(n) and key chords in context kcic(n), 1 < n < m, where m is the total number of pitches or chords in a section, are applied. kpic(n) and kcic(n) are modeled after *key words in context*, kwic(n), that helps show regular pitch and chord strings in the song.

A sample of kcic(2), n = 2, is shown in Fig. 6 (right) on the circle of fifths with their frequencies of occurrences. It is immediately obvious that *Despacito* has 4 prominent chords, D major, G major, A major and B minor. Chord progressions B minor to G major appears 22 times, G major-D major (18 times), D major-A major (18 times), and finally A major-B minor (10 times). While B minor-G major is most frequent, there is no G major-B minor progression.

Twenty-six song sections are displayed with frequencies of pitches on the music clefs and their proper positions, as well as their pitch classes on the circle of fifths, together with their music forms. Changes in the circles of fifths visualize the class content shifts in forms of song progression by sections.

H. Lyrics in phrase bars

In this final panel, for now, another way of data presentation can be viewed in a familiar manner, in a phrase

by phrase layout, containing lyrics with chords associated with their proper lyric syllables. This presentation is a shorthand way for viewers to sing along without being bothered with music clefs and music pitch notes, measures or beats.

Items J. to L. currently take about 20 seconds by the EMP, can be viewed at http://mlp.cs.nyu.edu/vietmusic/au2ana.php.

I. Current Issues

At the moment, the music processor adopts temporarily the use of the best spectrogram analysis by current leading software. This decision creates discrepancies in data units, and requires data unification into *milliseconds* and *cents*. There are other practical issues such as the best approaches for data presentation in layers, as the EMP for *Despacito* currently transfers a massive 330MB of graphic data to browsers.

The EMP begins to modestly scale up to over 4 minutes with 281,522x351 spectrogram of 99 million data points. The first spectrographic and all manual onsets of 9 layers took too long, 235 seconds of processing time. However, they are intended to be processed each separately according to the preference of viewers. The second analysis now takes 20 seconds to produce all 5 subparts, which can also be presented separately based on the viewer preference.

V. EDUCATIONAL IMPLICATIONS

The main goal of this paper is to work toward a model of music teaching and learning that would be more generally accessible, while not sacrificing the contents and quality of musical experiences.

The use of digital technologies to facilitate teachers and students to develop musical skills and knowledge, most efficiently in the following ways. First, the EMP provides a graphical analogue of music that bypasses the need for specialized musical notation and symbols. Graphical representation is especially suited for teaching musics whose tonalities do not often fit well with the western chromaticallybased tonal system. In this sense, these graphs help describing the music inputs faithfully with visual aids. After that, the EMP proceeds to calculate prevailing rhythmic and melodic patterns of a songs by constantly comparing units of data based on the theory of string grammar and deep learning. These patterns serve as basic blocks for improvisation and stylistic recognition. Finally, it allows any teachers to analyze and device lesson plans on any songs suggested by the students. The lessons are automatically tailored to match the student's ability. Teachers are then able to have more time focusing on accommodating other peculiar student needs, encouraging individual interpretations, and providing contextual knowledge.

ACKNOWLEDGMENT

The authors thank the *Linguistic String Project*'s Medical Language Processing of the Computer Science Department, New York University Courant Institute of Mathematical

Sciences, and the *Center for Vietnamese Philosophy, Culture & Society* of Temple University's Henry Luce Grant No. 270380 / *Expanding Opportunities for Research on Vietnam*, as well as Ambassador Nguyễn Phương Nga and the Cultural Section of Permanent Mission of the Socialist Republic of Vietnam to the United Nations who donated a top rank Dell workstation toward the EMP project.

REFERENCES

- [1] Z. S. Harris, String Analysis of Sentence Structure. The Hague: Mouton & Co., 1962.
- [2] P. Vĩnh, "Hệ thống Điệu và Hơi trong nghệ thuật ca đàn Huế," [A system of scales and airs in the arts of songs and music of Ca Hue], Chim Việt Cành Nam, 2014, retrieved at http://chimvie3.free.fr/54/vinhphuc cahue.htm.
- [3] T. N. Ngô and G. A. T Phan, "A Contribution to Teaching Vietnamese Music: Key Pitches in Context and Pitch Contour Graph," Journal of Social Sciences and Humanities, vol. 3, no. 5, pp. 573-585, Hanoi: Vietnam, 2017.
- [4] G. A. T. Phan and T. N. Ngô, "Capturing the Music: A case study of lý con sáo three regional "Songs of the Starling"," presented at the 46th Annual Mid-Atlantic Region Association for Asian Studies Conference: Mobility, Technology and the Environment, Drexel University, Philadelphia, PA, 2017.
- [5] G. A. T. Phan and T. N. Ngô, "Initial thoughts on analyzing sonic objects to aid multicultural education," presented at the 2017 Asian American Education Conference, Teachers College, Columbia University, New York City, NY, 2017.
- [6] N. Sager and T. N. Ngô, "The computability of strings, transformations, and sublanguage," The legacy of Zellig Harris: Language and information into the 21st Century, ed. by Bruce E Nevin and Stephen Johnson. John Benjamins Publishing Co. Amsterdam/Philadelphia. Volume 2:79-120, 2002.
- [7] Center for Digital Music of Queen Mary, University of London, Sonic Visualiser, version 3.1.1, 2018, retrieved at https://www.sonicvisualiser.org/.
- [8] M. Mauch and S. Dixon, "Approximate Note Transcription for the Improved Identification of Difficult Chords," proceedings of the 11th International Society for Music Information Retrieval Conference (ISMIR 2010), 2010.
- [9] J. Salamon and E. Gómez, "Melody Extraction from Polyphonic Music Signals using Pitch Contour Characteristics", IEEE Transactions on Audio, Speech and Language Processing, 20(6):1759-1770, Aug. 2012.
- [10] D. Barry, D. Fitzgerald, B. Lawlor, and E. Coyle, "Drum Source Separation using Percussive Feature Detection and Spectral Modulation." in ISSC 2005. 2005.
- [11] C. Baume and B. Y. Raimond, Broadcasting Corporation, Intensity Vamp plugin, 2013, retrieved at https://github.com/bbc/bbc-vampplugins/releases.
- [12] MuseScore BVBA, 2018, retrieved at https://musescore.org/en.
- [13] MuseScore BVBA, 2018, retrieved at https://musescore.com/dashboard.
- [14] Lunaverus, AnthemScore, 2018, retrieved at https://www.lunaverus.com/.
- [15] The Unicode Standard, version 11.0, 2018/06/05, retrieved at https://www.unicode.org/versions/Unicode11.0.0/.
- [16] World Wide Web, HyperText Markup Language (HTML) 5.2, 2017, retrieved at https://www.w3.org/TR/2017/REC-html52-20171214/.
- [17] World Wide Web, Extensible Markup Language (XML) 1.0 (Fifth Edition), 2013, retrieved at https://www.w3.org/TR/xml/.
- [18] Open Archive Initiative Object Reuse and Exchange (OAI/ORE), ORE User Guide - Primer, 2008, retrieved at https://www.openarchives.org/ore/1.0/primer.
- [19] Dublin Core Metadata Initiative, DCMI Specifications, 2005, retrieved at http://dublincore.org/specifications/.