Artificial Intelligence and Multimodal Signal Processing in Human-Machine Interaction



Series Editor: Abdulhamit Subasi **Volume Editors:** Abdulhamit Subasi, Saeed Mian Qaisar, and Humaira Nisar



ARTIFICIAL INTELLIGENCE AND MULTIMODAL SIGNAL PROCESSING IN HUMAN-MACHINE INTERACTION

Artificial Intelligence Applications in Healthcare and Medicine

ARTIFICIAL INTELLIGENCE AND MULTIMODAL SIGNAL PROCESSING IN HUMAN-MACHINE INTERACTION

Edited by

ABDULHAMIT SUBASI

University of Turku, Turku, Finland; Effat University, Jeddah, Saudi Arabia

SAEED MIAN QAISAR

CESI LINEACT, Lyon, France; College of Engineering, Effat University, Jeddah, Saudi Arabia

Humaira Nisar

Department of Electronic Engineering, Universiti Tunku Abdul Rahman, Kampar, Malaysia



Contents

Contributors ix	
Series preface xi	i
Preface xiii	
Acknowledgmen	ts xvii

Introduction to human-machine interaction

SYED SAAD AHMED, HUMAIRA NISAR, AND PO KIM LO

- 1.1 Introduction 1
- 1.2 HMI in the workplace 3
- 1.3 HMI in various fields 8
- 1.4 The human component of HMI 13
- 1.5 Conclusion 15 References 16
 - Artificial intelligence techniques for human-machine interaction

HAMID MUKHTAR

- 2.1 Introduction 19
- 2.2 Artificial intelligence techniques for HMI 26
- 2.3 Conclusions and future directions 39 References 39
 - 3. Feature extraction techniques for human-computer interaction

ABDULHAMIT SUBASI AND SAEED MIAN QAISAR

- 3.1 Introduction 43
- 3.2 The spectral analysis 46
- 3.3 The time-frequency analysis 50
- 3.4 Common spatial patterns (CSP) 58
- 3.5 Conclusion 59

References 59

An overview of techniques and best practices to create intuitive and user-friendly human-machine interfaces

VEERENDRA DAKULAGI, KIM HO YEAP, HUMAIRA NISAR, ROHINI DAKULAGI, G N BASAVARAJ, AND MIGUEL VILLAGÓMEZ GALINDO

- 4.1 Introduction to human-machine interaction 63
- 4.2 Technological foundations for intuitive interfaces 68
- 4.3 Multimodal signal processing and interface integration 70
- 4.4 Best practices in interface design 71
- 4.5 Future trajectory of intuitive interfaces 72
- 4.6 Ethical considerations in interface influence 73
- 4.7 HMI technologies in different industries 74
- 4.8 Conclusion 76 References 77

An overview of electroencephalogram based human-computer interface

MD MAHMUDUL HASAN, SITI ARMIZA MOHD ARIS, AND NORIZAM SULAIMAN

- 5.1 Introduction 79
- 5.2 Fundamentals of EEG 81
- EEG, electrode placement systems, and configurations 81
- 5.4 EEG data acquisition 83
- 5.5 EEG-based HCI components 84
- Innovations and challenges in EEG-based HCI 90
- 5.7 Future directions and emerging trends 92
- 5.8 Conclusion 93

References 94

vi

Contents

Speech-driven human-machine interaction using Mel-frequency Cepstral coefficients with machine learning and Cymatics Display

SAEED MIAN OAISAR

6.1 Introduction 103

6.2 Materials and methods 107

6.3 System implementation and results 116

6.4 Conclusion 121

References 122

 EEG-based brain-computer interface using wavelet packet decomposition and ensemble classifiers

ABDULHAMIT SUBASI AND SAEED MIAN QAISAR

- 7.1 Introduction 125
- 7.2 Literature review 127
- 7.3 Materials and methods 127
- 7.4 Results and discussion 134
- 7.5 Conclusion 147

Acknowledgment 148

References 148

 Understanding dyslexia and the potential of artificial intelligence in detecting neurocognitive impairment in dyslexia

SITI ATIYAH ALI, HUMAIRA NISAR, NURFAIZATUL AISYAH AB AZIZ, NOR ASYIKIN FADZIL, NUR SAIDA MOHAMAD ZABER. AND LUTHEFI IDZHAR ISMAIL

- 8.1 Introduction to dyslexia and artificial intelligence 151
- 8.2 A brief introduction: Learning process of normal children vs dyslexic children 153
- 8.3 Electroencephalography 155
- 8.4 ERP in dyslexia 157
- 8.5 The future of dyslexia: AI assistance in learning for dyslexia 165
- 8.6 The future of dyslexia: ML for early diagnosis 166
- 8.7 Conclusion 167

Acknowledgment 167 Declaration 168 References 168

> Early dementia detection and severity classification with deep SqueezeNet convolutional neural network using EEG images

NOOR KAMAL AL-QAZZAZ, SAWAL HAMID BIN MOHD ALI, AND SITI ANOM AHMAD

- 9.1 Introduction 171
- 9.2 Related work 172
- 9.3 Materials and methods 173
- 9.4 Results and discussion 178
- 9.5 Conclusions 180

References 182

EEG-based stress identification using oscillatory mode decomposition and artificial neural network

SARIKA KHANDELWAL, NILIMA SALANKAR, AND SAEED MIAN QAISAR

10.1 Introduction 185

10.2 Materials and adopted methodology 189

10.3 Results and discussion 194

10.4 Conclusion 206

Acknowledgment 207

References 207

11. EEG signal processing with deep learning for alcoholism detection

HAMID MUKHTAR

11.1 Introduction 211

- 11.2 Materials and methods in EEG signal processing 214
- 11.3 Deep neural networks for EEG analysis 219
- 11.4 Discussion 222
- 11.5 Conclusions and future of deep learning in alcoholism detection 224

References 225

Contributors

- Abhishek Department of Radiology, Kalaish Institute of nursing and Paramedical sciences. Greater Noida, Uttar Pradesh, India
- Siti Anom Ahmad Department of Electrical and Electronic Engineering, Faculty of Engineering, Universiti Putra Malaysia, UPM; Malaysian Research Institute of Ageing (MyAgeing), Universiti Putra Malaysia, Serdang, Selangor, Malaysia
- Syed Saad Ahmed Department of Electronic Engineering, Faculty of Engineering and Green Technology, Universiti Tunku Abdul Rahman, Kampar, Perak; Centre for Healthcare Science and Technology, Universiti Tunku Abdul Rahman, Sungai Long, Selangor, Malaysia
- Sawal Hamid Bin Mohd Ali Department of Electrical, Electronic & Systems Engineering, Faculty of Engineering & Built Environment, Universiti Kebangsaan Malaysia, UKM, Bangi; Department of Electrical, Electronic and Systems Engineering, Centre of Advanced Electronic and Communication Engineering, Universiti Kebangsaan Malaysia, Selangor, Malaysia
- Siti Atiyah Ali Department of Electronic Engineering, Faculty of Engineering and Green Technology, Universiti Tunku Abdul Rahman, Kampar, Perak; Centre for Healthcare Science and Technology, Universiti Tunku Abdul Rahman, Sungai Long, Selangor; Faculty of Cognitive Sciences and Human Development, Universiti Malaysia Sarawak, Kota Samarahan, Sarawak, Malaysia
- Noor Kamal Al-Qazzaz Department of Biomedical Engineering, Al-Khwarizmi College of Engineering, University of Baghdad, Baghdad, Iraq
- Nurfaizatul Aisyah Ab Aziz Faculty of Cognitive Sciences and Human Development, Universiti Malaysia Sarawak, Kota Samarahan, Sarawak, Malaysia

- G N Basavaraj Department of ISE, BMS Institute of Technology and Management Bangalore, India
- Ayse Kosal Bulbul Faculty of Medicine, Institute of Biomedicine, University of Turku, Turku, Finland
- Amit Pratap Singh Chouhan Department of Radiology, Sharda School of Allied Health Sciences, Sharda University, Greater Noida, Uttar Pradesh, India
- Rohini Dakulagi Department of Electrical and Electronics Engineering, Guru Nanak Dev Engineering College, Bidar, Karnataka, India
- Veerendra Dakulagi Department of CSE (Data Science), Guru Nanak Dev Engineering College, Bidar, Karnataka, India
- Nor Asyikin Fadzil Department of Psychiatry, School of Medical Sciences, Universiti Sains Malaysia, Kota Bharu, Kelantan, Malaysia
- Francisco Ferrero Department of Electrical, Electronic, Communications, and Systems Engineering, University of Oviedo, Gijón, Spain
- Miguel Villagómez Galindo Michoacan University of San Nicolás de Hidalgo, Morelia, Mexico
- Andrea Guazzini Department of Education, Literatures, Intercultural Studies, Languages and Psychology, University of Florence, Florence, Italy
- Mustafa Can Gursesli Department of Information Engineering, University of Florence, Florence, Italy
- Md Mahmudul Hasan Faculty of Electrical and Electronics Engineering Technology, Universiti Malaysia Pahang Al-Sultan Abdullah, Pekan, Pahang, Malaysia
- Luthffi Idzhar Ismail Faculty of Engineering, Universiti Putra Malaysia, Serdang, Selangor, Malaysia
- Sarika Khandelwal G H Raisoni College of Engineering, Nagpur, India

Understanding dyslexia and the potential of artificial intelligence in detecting neurocognitive impairment in dyslexia

Siti Atiyah Ali^{a,b,c}, Humaira Nisar^{a,b}, Nurfaizatul Aisyah Ab Aziz^c, Nor Asyikin Fadzil^d, Nur Saida Mohamad Zaber^a, and Luthffi Idzhar Ismail^e

^aDepartment of Electronic Engineering, Faculty of Engineering and Green Technology, Universiti Tunku Abdul Rahman, Kampar, Perak, Malaysia ^bCentre for Healthcare Science and Technology, Universiti Tunku Abdul Rahman, Sungai Long, Selangor, Malaysia ^cFaculty of Cognitive Sciences and Human Development, Universiti Malaysia Sarawak, Kota Samarahan, Sarawak, Malaysia ^dDepartment of Psychiatry, School of Medical Sciences, Universiti Sains Malaysia, Kota Bharu, Kelantan, Malaysia ^cFaculty of Engineering, Universiti Putra Malaysia, Serdang, Selangor, Malaysia

8.1 Introduction to dyslexia and artificial intelligence

In 1877 Dr. Adolph Kussmaul, a German physician, introduced the term "word blindness" to describe a condition of individuals who had the ability to distinguish the form of letters but were unable to read and translate them into spoken words. A decade later, a German ophthalmologist, Dr. Rudolf Berlin, coined the term "dyslexia," which derived from the Greek words dys- (difficult) and -lexia (words), and literally means "difficulty with words." Reading is not an easy task. Most of us mastered our reading abilities during the early years of primary

school, where we learned how to pronounce each letter and connect the sound of each letter to form words. Unfortunately, this ability is not present in children with dyslexia. Dyslexia is a neurodevelopmental disorder that is referred to as a specific learning disorder (SLD) in reading and writing, which affects academic performance. While dyslexia can also arise later in life from head trauma, illness, or other factors, the most common type is present at birth and persists throughout life. Most often, children with dyslexia have delayed learning development compared to their peers of the same age. It is not new knowledge that this difficulty originally comes from neurocognitive abnormalities,