Indexing Cerebrovascular Health Using Near-Infrared Spectroscopy: A Multi-Model Analysis

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A thesis submitted in fulfilment of the requirements for the degree of

Doctor of Philosophy

School of Electrical Engineering and Computing The University of Newcastle, Australia

July 2020



This research is supported by an Australian Government Research Training Program (RTP) Scholarship

Statement of Originality

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Rashid Ghorbani Afkhami

Abstract

Cerebrovascular health is of great concern, especially in the ageing population, as stiff vessels are linked with diseases such as stroke, dementia, and age-related disabilities. Data from brain imaging techniques, including magnetic resonance imaging (MRI) and transcranial Doppler ultrasound (TCD), can be used to infer information regarding the health of underlying arteries. Such health assessments are commonly based on an index that is dependent on the properties of the device and can be related to known vascular health factors by a mathematical model.

The contribution of this thesis to the field of cerebrovascular health and brain imaging is twofold: First, we develop a timing index (TI) as a measure of cerebrovascular health. The relationship between TI and vascular health factors is derived in the context of pressure waveforms using transmission line theory and Windkessel model models. The proposed formula for TI matches data reported in the literature and helps to understand the flattening behaviour observed in the arrival time of reflected waves in aged subjects. Using similar mathematical modelling we also derive an expression for the relationship between the existing augmentation index (AI) and the same vascular health factors. Together with existing pulsatility index (PI) results, we show mathematically that TI is potentially more strongly related to vessel stiffness than either of the two indices currently used to index cerebrovascular health i.e., PI and AI. This is particularly so in younger to middle-aged subjects where interventions are best applied. We then show that TI can be applied to TCD measurements of blood flow velocity. To our knowledge, this is the first use of wave reflection time to measure vascular health in the brain. Transcranial Doppler Ultrasound Timing Index (TI_{TCD}) shows a significant correlation with age. Furthermore, compared to the existing transcranial Doppler augmentation index (AI_{TCD}) and transcranial Doppler pulsatility index (PI_{TCD}), the TI_{TCD} show stronger correlations with cardiorespiratory fitness and the magnetic resonance imaging pulsatility index (PI_{MRI}) .

The second contribution of this thesis is in its application of near-infrared spectroscopy (NiRS). Firstly, we propose a NiRS signal model capable of producing synthetic NiRS signals comprising low-frequency components, arterial pulsation signals, reflected waves, Mayer and respiratory waves and a haemodynamic response function. The model outputs are compared with measured NiRS signals, and it is shown that the modelled signal is equivalent to the recorded signal as a later set of recordings on the same channel. Then, as an emerging tool for measuring cerebrovascular health, we propose a novel algorithm for locating systolic and reflected peaks on an averaged NiRS signal, thereby applying the TI to NiRS. The new near-infrared spectroscopy timing index (TI_{NiRS})

shows stronger correlations with age, cardiorespiratory fitness (CRF) and PI_{MRI} than the pulse relaxation function (PReFx) which is an existing NiRS-based vascular health index.

Compared with existing brain imaging techniques, NiRS offers several advantages, such as being inexpensive, portable and easy-to-use. The NiRS-related contributions of this thesis are the development of a NiRS signal model and a NiRS-based cerebrovascular health measure. These will help in the development of a technique for the routine clinical measurement of cerebrovascular health. Such a technique would facilitate early intervention in the progression of vascular stiffness with age and, potentially, vascular-related diseases such as stroke and dementia.

Acknowledgements

First, I would like to thank my supervisor, Professor Sarah Johnson, for her continuous support and guidance throughout my PhD. Her availability despite a busy schedule and her comments and brilliant ideas made this thesis possible.

I would like to extend my thanks to my co-supervisors Professor Rohan Walker and Associate Professor Saadallah Ramadan for their support and insight. Special thanks to Dr Rachel Wong for her detailed comments on the papers.

It should be mentioned that this thesis would not have been possible without the volunteers who participated in the study. I must also express my appreciation to Hamish Evans, Shiami Luchow, Kristen Fisher and Jillian Mason for their help in data collection and recruitment.

Last but not least, I thank my wife Safa for her unconditional love and support during my PhD and the sacrifices she made to be with me on this journey. I also want to thank our family and all our friends who made this wonderful experience in Australia possible for us.

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Abbreviations

AHA American Heart Association

AI Augmentation Index

AI_{TCD} Transcranial Doppler Augmentation Index
AI* Augmentation Index, Alternative Definition

AP Arterial PulsationASL Arterial Spin Labelling

ba-PWV Brachial-ankle Pulse Wave Velocity
 bf-PWV Brachial-femoral Pulse Wave Velocity
 BOLD Blood-oxygenation Level Dependent

bpm Beats Per MinuteBR Breathing Rate

CBF Cerebral Blood Flow

cf-PWV Carotid-femoral Pulse Wave Velocity

CFEn Cross-fuzzy Entropy

CPP Cerebral Perfusion PressureCRF Cardiorespiratory FitnessCSF Cerebrospinal Fluid

CVR Cerebrovascular Resistance

DC Direct Current

DCS Diffuse Correlation Spectroscopy

ECG Electrocardiography
EEG Electroencephalography

fMRI Functional Magnetic Resonance Imaging

HR Heart Rate

HRF Haemodynamic Response Function

ICP Intracranial PressureIIR Infinite Impulse Response

LF Low-frequency

MAPMean Arterial PressureMBPMean Blood PressureMCAMiddle Cerebral Artery

Abbreviations xii

MRI Magnetic Resonance Imaging

NiR Near-infrared

NiRS Near-infrared Spectroscopy

PC Phase Contrast PI Pulsatility Index

 $\mathbf{PI_{MRI}}$ Magnetic Resonance Imaging Pulsatility Index

PI_{TCD} Transcranial Doppler Pulsatility Index

PI*_{TCD} Transcranial Doppler Pulsatility Index, Alternative

Definition

PP Pulse Pressure

PReFxPulse Relaxation FunctionPWVPulse Wave Velocity

SNR Signal-to-noise Ratio
 std Standard Deviation
 SVD Small Vessel Disease

 $\mathbf{T_{refl}}$ Reflection Time

TCD Transcranial Doppler Ultrasound

TI Timing Index

 $\begin{array}{ll} \mathbf{TI_{NiRS}} & \text{Near-infrared Spectroscopy Timing Index} \\ \mathbf{TI_{TCD}} & \text{Transcranial Doppler Ultrasound Timing Index} \end{array}$

TL Transmission Line
TOST Two One-sided T-tests
TPR Total Peripheral Resistance

VLF Very Low Frequency

WGN White Gaussian Noise WK Windkessel Model

WK2 Two-element Windkessel Model
 WK3 Three-element Windkessel Model
 WK4 Four-element Windkessel Model

Symbols

| Symbol | Definition | Units |
|---------------|------------------------|---|
| C | Compliance | m^3/Pa |
| P | Pressure | Pa (1 Pa \approx 1/133 mmHg, or N/m²) |
| Q | Flow | $m^3/s (1m^3/s = 10^3l/s)$ |
| R | Resistance | $\mathrm{Pa/m^3}$ |
| V | Velocity | m/s |
| \mathcal{V} | Volume | $m^3 (1m^3 = 10^3 l)$ |
| Z | Impedance | $\mathrm{Pa/m^3}$ |
| β | Phase Constant | rad/m |
| Γ | Reflection Coefficient | unitless |
| γ | Propagation Constant | rad/m |
| ω | Angular Velocity | rad/s |