

UNIVERSITI PUTRA MALAYSIA

EXPLORING NEURONAL ACTIVATION IN RESPONSE TO REWARD THROUGH MRI CORRELATED WITH SIMULTANEOUS EEG AMONG MALE UNDERGRADUATES HAVING PROBLEMATIC SMARTPHONE USAGE

SHARIFAT HAMED

FPSK(m) 2021 28



EXPLORING NEURONAL ACTIVATION IN RESPONSE TO REWARD THROUGH fMRI CORRELATED WITH SIMULTANEOUS EEG AMONG MALE UNDERGRADUATES HAVING PROBLEMATIC SMARTPHONE USAGE

By

SHARIFAT HAMED

Thesis Submitted to the School of Graduate Studies, Universiti Putra Malaysia, in Fulfilment of the Requirements for the Degree of Master of Science

August 2020

COPYRIGHT

All material contained within the thesis, including without limitation text, logos, icons, photographs, and all other artwork, is copyright material of Universiti Putra Malaysia unless otherwise stated. Use may be made of any material contained within the thesis for non-commercial purposes from the copyright holder. Commercial use of material may only be made with the express, prior, written permission of Universiti Putra Malaysia.

Copyright © Universiti Putra Malaysia



Abstract of thesis presented to the Senate of Universiti Putr Malaysia in fulfilment of the requirement for the degree of Master of Science

EXPLORING NEURONAL ACTIVATION IN RESPONSE TO REWARD THROUGH fMRI CORRELATED WITH SIMULTANEOUS EEG AMONG MALE UNDERGRADUATES HAVING PROBLEMATIC SMARTPHONE USAGE

By

SHARIFAT HAMED

August 2020

Chairman : Associate Professor Subapriya Suppiah, PhD, MD, MMed Faculty : Medicine and Health Sciences

Smartphone usage among young Malaysian adults has increased dramatically in the past decade leading to smartphone addiction (SPA) particularly involving social networking applications such as Instagram. SPA prevalence has been noted to be as high as 47% among medical students in a Malaysian university. There is a need to objectively assess the cerebral mechanisms that respond to reward using electroencephalography (EEG)-informed functional magnetic resonance imaging (fMRI) also known as EEG-fMRI as a potential biomarker.

Therefore, after receiving ethical clearance, an observational study was conducted among 24 male UPM students using Smartphone Addiction Scale-Malay version (SAS-M) questionnaire, modified Instagram Addiction Scale (IAS) and EEG-fMRI. The subjects were recruited by simple random sampling from a Phase 1 cross sectional study comprising of 850 male UPM students who had been administered with the SAS-M questionnaire. Subjects with SAS-M scores \geq 98 and IAS \geq 37 were considered High scorers (HS) and deemed to be having SPA, whereas subjects with SAS-M scores < 98 and IAS < 37 were considered Low Scorers (LS) and deemed as healthy controls. A 64-channel EEG scalp electrode was placed on the participants (12 HS, 12 LS) and a task-based fMRI was performed with simultaneous EEG recordings.

Evoked response potential (ERP) derivatives of EEG namely the P300 peak waves and contingent negativity variance (CNV) were analyzed using Brain-Vision EEG analyzer and fMRI dataset were analyzed using Statistical Parametric Mapping (SPM). There was significant difference in the P300 wave amplitude between HS and LS, which corresponded well with cerebral activations regions related to response to reward.

The combination of the high spatial resolution of fMRI with the high temporal resolution of EEG to correlate cerebral regional activation, in response to cue related reactivity during response to reward task among smartphone addicts, has the potential to be a surrogate biological marker for assessment of Instagram addiction.



Abstrak tesis yang dikemukakan kepada Senat Universiti Putra Malaysia sebagai memenuhi keperluan untuk Ijazah Master Sains

KORELASI PENGAKTIFAN NEURONAL MENGGUNAKAN fMRI DALAM TINDAKBALAS TERHADAP GANJARAN DENGAN EEG-SERENTAK DI KALANGAN MAHASISWA YANG MENGALAMI PENGGUNAAN TELEFON PINTAR BERMASALAH

Oleh

SHARIFAT HAMED

Ogos 2020

Pengerusi : Profesor Madya Subapriya Suppiah, PhD, MD, MMed Fakulti : Perubatan dan Sains Kesihatan

Penggunaan telefon pintar di kalangan generasi muda di Malaysia telah meningkat baru-baru ini dan mengakibatkan ketagihan telefon pintar ataupun penggunaan telefon pintar bermasalah (SPA) terutama yang melibatkan aplikasi rangkaian sosial seperti Instagram. SPA dijangka mempengaruhi hampir 47% di kalangan pelajar perubatan di salah sebuah universiti awam di Malaysia. Terdapat keperluan untuk mengkaji secara objektif untuk menilai mekanisma otak yang bertindakbalas terhadap ganjaran menggunakan pengimejan resonans magnetik yang berfungsi (fMRI) serta elektroencephalographi (EEG) serentak, yang juga dikenali sebagai EEG-fMRI sebagai biomarker yang berpotensi.

Setelah menerima kelulusan etika, kajian observasi telah dijalankan di kalangan 24 pelajar siswa UPM menggunakan borang soal selidik versi Bahasa Melayu - Skor Ketagihan Smartphone-Melayu (SAS-M) serta Skala Instagram Addiction yang dimodifikasi (IAS) dan EEG-fMRI. Para subjek dijemput untuk kajian Fasa ke-2 melalui kaedah pensampelan mudah rawak dari Kajian Fasa Pertama yang terdiri daripada 850 pelajar siswa ijazah UPM yang menjawab boring kajiselidik SAS-M. Subjek-subjek dengan skor SAS-M ≥ 98 serta IAS ≥ 37 dianggap Penilai Tinggi atau 'High Scorer' (HS) dan dianggap mempunyai SPA, manakala subjek dengan skor SAS-M <98 serta IAS < 37 dianggap 'Low Scorer' (LS) dan dianggap sebagai subjek kawalan yang sihat. Elektrod kulit kepala EEG yang mempunyai 64 saluran diletakkan pada 24 peserta (12 HS, 12 LS) dan fMRI berasaskan tugasan melihat imej yang dipapar di skrin dilakukan dengan rakaman EEG serentak.

Derivatif potensi respons yang teruja (ERP) terutamanya gelombang P300 dan variasi contigen negatif (CNV) telah dianalisa dengan menggunakan penganalisis EEG Brain-Vision dan dataset fMRI dianalisa menggunakan Pemetaan Statistik Parametrik (SPM). Terdapat perbezaan yang signifikan dalam ERP yang ditimbulkan pada amplitud gelombang P300 antara HS dan LS; yang sesuai dengan pengaktifan serebrum wilayah yang berkaitan dengan tindak balas terhadap ganjaran. Gabungan resolusi spatial tinggi fMRI dengan resolusi temporal yang tinggi EEG untuk mengaitkan pengaktifan serantau, sebagai tindakbalas kepada rangsangan berkaitan reaksi semasa tindakbalas kepada tugasan ganjaran di kalangan pelajar dengan SPA, berpotensi menjadi penanda biologi tumpuan untuk penilaian ketagihan telefon pintar, khususnya kepada Instagram.



ACKNOWLEDGEMENTS

I offer my most sincere gratitude to my extraordinary supervisor, Associate Professor Dr. Subapriya Suppiah for her invaluable guidance and constant encouragement. I would also like to express my gratitude to my supervisory committee members, Dr. Nur Farhayu Omar and Professor Madya Dr. Ezamin Abdul Rahim, for their insightful discussions and encouragement.

I would like to thank several advisors of this project in their advisory role and guidance regarding this research project. Firstly, I would like to thank Dr. Mazlyfarina Mohamad from the Department of Diagnostic & Applied Health Sciences, Universiti Kebangsaan Malaysia (UKM) for sharing her expertise in SPM fMRI analysis and Dr. Rohit Tyagi scientific consultant at Aerobe Pte Lte Singapore for sharing his expertise in the fMRI paradigm preparation and EEG software analysis. I would also like to thank Dr Beatrice Ng Andrew (Department of Psychiatry), Associate Professor Dr Ching Siew Mooi (Family Medicine Physician with special interest in Psychiatry and Young Adult Health) and Associate Professor Dr Hoo Fan Kee (Neurology Unit, Department of Internal Medicine), Faculty of Medicine and Health Sciences, Universiti Putra Malaysia for their guidance and advice related to addiction, preparation of methodology and recruitment of study subjects.

I wish to express my gratitude to all the members at the Centre for Diagnostic Nuclear Imaging (PPDN), Universiti Putra Malaysia for their assistance in my research project. My appreciation is also extended to the radiographers for their assistance in scanning the research subjects. I also wish to express my gratitude to my colleagues Ms. Nisha Syed Nasser and Dr. Aida Abdul Rashid for the teamwork during this research project. I would also like to acknowledge the financial support provided by the Faculty of Medicine & Health Sciences Universiti Putra Malaysia (UPM) and the Research Management Centre, Universiti Putra Malaysia that awarded the research grant for this project i.e. Geran Putra IPS (GP-IPS/2017/9580800) and Geran Putra (GP/2017/9549800).

A very special appreciation to all the undergraduate students in UPM, who managed their class schedule well in order to take part in this research project and helped to finish this project on time.

Last but not the least, I would like to thank my family: my parents and my brothers and sister for supporting me spiritually throughout this project, writing this thesis and my life in general.

This thesis was submitted to the Senate of Universiti Putra Malaysia and has been accepted as fulfilment of the requirement for the degree of Master of Science. The members of Supervisory Committee were as follows:

Subapriya Suppiah, PhD, MD, MMed

Associate Professor Faculty of Medicine and Health Sciences Universiti Putra Malaysia (Chairman)

Nur Farhayu Binti Omar, PhD

Senior Lecturer Faculty of Medicine and Health Sciences Universiti Putra Malaysia (Member)

Ezamin Bin Abdul Rahim, PhD, MD, MMed

Professor Faculty of Medicine and Health Sciences Universiti Putra Malaysia (Member)

ZALILAH MOHD SHARIFF, PhD

Professor and Dean School of Graduate Studies Universiti Putra Malaysia

Date: 11 March 2021

Declaration by graduate student

I hereby confirm that:

- this thesis is my original work;
- quotations, illustrations and citations have been duly referenced;
- this thesis has not been submitted previously or concurrently for any other degree at any institutions;
- intellectual property from the thesis and copyright of thesis are fully-owned by Universiti Putra Malaysia, as according to the Universiti Putra Malaysia (Research) Rules 2012;
- written permission must be obtained from supervisor and the office of Deputy Vice-Chancellor (Research and innovation) before thesis is published (in the form of written, printed or in electronic form) including books, journals, modules, proceedings, popular writings, seminar papers, manuscripts, posters, reports, lecture notes, learning modules or any other materials as stated in the Universiti Putra Malaysia (Research) Rules 2012;
- there is no plagiarism or data falsification/fabrication in the thesis, and scholarly integrity is upheld as according to the Universiti Putra Malaysia (Graduate Studies) Rules 2003 (Revision 2012-2013) and the Universiti Putra Malaysia (Research) Rules 2012. The thesis has undergone plagiarism detection software

0:	
Signatur	e:
orginatai	۰.

Date:

Name and Matric No: Sharifat Hamed, GS51067

Declaration by Members of Supervisory Committee

This is to confirm that:

- the research conducted and the writing of this thesis was under our supervision;
- supervision responsibilities as stated in the Universiti Putra Malaysia (Graduate Studies) Rules 2003 (Revision 2012-2013) were adhered to.

Signature: Name of Chairman of Supervisory Committee:	Associate Professor Dr. Subapriya Suppiah
Signature: Name of Member of Supervisory Committee:	Dr. Nur Farhayu Binti Omar
Signature: Name of Member of Supervisory Committee:	Professor Dr. Ezamin Bin Abdul Rahim

TABLE OF CONTENTS

			Page
ABSTRAC ABSTRAF ACKNOW APPROVA DECLARA LIST OF T LIST OF A LIST OF A	CT (LEDGE AL ATION CABLES FIGURES APPEND ABBREV	MENTS S ICES IATIONS	i iii v vi viii xiii xv xvii xviii
CHAFTER	·		
1	INTRO 1.1 1.2 1.3 Genera 1.4 1.5 1.6	DUCTION Background of Study Problem Statement and Justification Objectives I Objective: Hypothesis Organization of Thesis Conceptual Framework	1 1 3 5 5 5 7 8
2			0
2	2.1	Smartphone Introduction	9
	2.2	Smartphone Usage in Malaysia	9
	2.3	Smartphone Addiction and Its Social Effects	10
	2.4	Questionnaire Evolution of Smartphone Addiction	14
		2.4.1 Smartphone Addiction Scale	14
		2.4.2 Modified Instagram Addiction (IAS)	15
	2.5	Functional Magnetic Resonance Imaging	15
	2.6	Reward System in Human Brain	16
	2.7	Lask-Based IMRI In Addiction	17
	2.0	Electroencephalography (EEG)	10
	2.9		19
		2.9.2 LORETA in EEG	20
	2.10	EEG Derivatives	22
	2.1.0	2.10.1 Event-related Potential (ERP)	22
		2.10.2 Contingent Negative Variation	23
		2.10.3 Evoked Potential	25
	2.11	fMRI BOLD-Imaging with Simultaneous-EEG	26

3	METH	ODOLOGY	35
	3.1	Study Design and Method	35
	3.2	Study Population	36
	3.3	Sample Size	37
	3.4	Ethics of Study	38
	3.5	Research Instruments	39
		3.5.1 Smartphone Addiction Scale-Malay Vers	sion
		(SAS-M)	39
		3.5.2 Modified Instagram Addiction Scale	40
	3.6	Data Acquisition and Imaging Protocol	40
		3.6.1 Structural fMRI Data Acquisition	41
		3.6.2 Functional MRI Data Acquisition	41
		3.6.3 EEG Data Acquisition	41
		3.6.4 EEG-correlated fMRI Technical Limitation	42
	3.7	Simultaneous fMRI-EEG Paradigms	42
	3.8	fMRI and EEG Data Processing and Analysis	43
		3.8.1 fMRI Data Processing	43
		3.8.2 Statistical Analysis	45
		3.8.3 EEG Data Processing	46
		3.8.4 Managing EEG Files in Analyzer Workspace	46
		3.8.5 Brain-Vision Software Processing Steps	48
		3.8.6 Time Markers (Import marker)	52
	3.9	Methodology Flowchart	53
4	RESU	TS	54
	4.1	Socio-demographic Response Rate	54
		4.1.1 Socio-demographic Factors of Respondents	54
		4.1.2 Pattern of Smartphone Usage am	ong
		Respondents	54
		4.1.3 Effects of Smartphone Usage	57
	4.2	EEG-fMRI Response Rate	58
	4.3	Characteristics of the Participants	58
		4.3.1 Demographic Data of the Participants Involved	IN
			58
		4.3.2 Tests of Normality	60
	4.4	Lask-based fMRI Findings	62
	4.5	ROI Analysis	65
	4.6		68
		4.6.1 ERP derivatives	69
		4.6.2 P300 Peak Wave	69
	. –	4.6.3 Contingent Negative Variation	73
	4.7	Correlation of results (Questionnaires, fMRI and EEG)	75
		4.7.1 Pearson's Correlation	75
Б	Discus	scion	70
5	DISCUS	221011	10
6	CONC	LUSION, LIMITATION OF STUDY AND	
	RECO	MMENDATION	86
	6.1	Conclusion	86
	6.2	Limitation	87
	0		5.

6.3	Recommendation for Future Research	87
REFERENCES APPENDICES BIODATA OF S LIST OF PUBL	STUDENT ICATIONS	88 101 112 113



(G)

LIST OF TABLES

Table		Page
2.1	Smartphone addiction measurement instruments devised until 2015	12
2.2	Indicates the results and study from 2015-2018 related to "Non- substance Addiction (Internet Addiction, Gambling addiction, Smartphone and Social-networks addiction) using Task-Based functional MRI" as there are very little studies about addiction using fMRI with simultaneous-EEG	28
2.3	Recent EEG-fMRI studies pertaining to meso-corticolimbic disorders	30
4.1	Distribution of socio-demographic factors of respondents (n=850)	54
4.2	The pattern of smartphone usage among respondents based on gender	55
4.3	Frequency distribution regarding the main purpose for using smartphone among students	55
4.4	Frequency distribution of most widely used Smartphone application	56
4.5	Distribution of participants according to being healthy or PIGU	57
4.6	Demographic data of 24 participants	59
4.7	Demographic parameter of 24 participants	59
4.8	Test of Normality for demographic data, smartphone and Instagram variables and EEG derivatives	60
4.9	Skewness and Kurtosis Tests of Normality	60
4.10	Test for Normality of PSC at selected ROI for 'Risky' pictures in (HS group > LS group)	61
4.11	Skewness and Kurtosis Tests for Normality of 'Risky' pictures in (HS group > LS group)	61
4.12	Whole Brain Activation in PIGU group in response to 'Risky' pictures	62

6

4.13	Whole Brain Activation in LS group in response to 'Risky' pictures	64
4.14	Whole Brain Activation in response to 'Risky' pictures in PIGU group as compared to Control group (PIGU group >control group)	65
4.15	Percentage signal change at selected ROI in response to 'Risky' pictures in HS group as compared to LS	66
4.16	ERP (P300 peak wave) at five central EEG channels, in "HS Risky vs. HS Participant" conditions	70
4.17	ERP (P300 peak wave) at five central EEG channels, in "Risky vs. Participant" conditions in Low Scorers	71
4.18	ERP (P300 peak wave) at five central EEG channels, in HS Risky vs. LS in 'Risky' conditions	73
4.19	CNV time detection of HS vs LS in participant condition at CP1 channel	74
4.20	CNV time detection of HS vs LS in 'Risky' condition at CP1 channel	74
4.21	Correlation between SAS-M and IAS score over fMRI ROI in 'Risky' Condition in High Scorer group	76
4.22	Correlation between SAS-M and Precuneus activation over EEG results (P300 and CNV) In 'Risky' Condition	76

C

LIST OF FIGURES

Figure	•	Page
1.1	The conceptual framework for simultaneous EEG-fMRI response to reward study	8
2.1	Raw EEG of an awake person (blue) and propofol-anesthetized person (red)	20
2.2	Filtered EEGs (<1.5 Hz) of an awake person (blue) and propofol-anesthetized person (red) (left panel) and corresponding FFT spectra (right)	21
2.3	The graph shows the ERP (P300 peak wave) signal that results from a target stimulus (PIGU group) versus signal from a non-target stimulus (Control group)	23
2.4	Contingent Negative Variation is a slow negative wave that develops in the interval between a "Warning" and a "Go" stimulus. The graph indicates the anticipation for a forthcoming signal and preparation for execution of a response	24
2.5	The graph displays Contingent Negative Variation (CNV) with early and late wave	24
2.6	Typical example of an evoked potential obtained from processing a Sender's signals. The graph is an average of 128 stimuli and 14 EEG channels. Usually two peaks are seen, a negative and a positive one, about 250 to 300 ms after the stimulus begins, and a minor peak at about 250 ms after the stimulus ceases	25
2.7	A sample of full field and half field "Pattern reversal visual evoked potentials" in a patient with multiple sclerosis. The pathway disorder is revealed only in the response to stimulation of the left half field	26
3.1	Visual Perception and Decision-making block design paradigms that have been used in this study. Each flash card displayed 6 second with total of 30 second in each 8 blocks	43
3.2	Overview of Statistical parametric map (SPM) steps	45
3.3	Analyzer button and main menu of the application	47
3.4	Opened raw EEG data file	47

3.5	flowchart of EEG data analysis using Brain-Vision software	49
3.6	CB correction, Pulse artifact correction	51
3.7	EEG data set sample after cleaning-up the artifacts, using brain vision software	52
3.8	Methodology flowchart for study protocol	53
4.1	Distribution of purpose of using smartphones among UPM students	56
4.2	Distribution of frequently used smartphone applications among UPM students	57
4.3	Superior, inferior, right and left of whole brain activation in HS group during 'Risky' Condition	63
4.4	Superior, inferior, right and left of whole brain activation in LS group during 'Risky' Condition	64
4.5	PSC comparison of HS vs LS in right dorsolateral prefrontal (RDLPF)	67
4.6	PSC comparison of HS vs LS in Left Precuneus	67
4.7	PSC comparison of HS vs LS in Right medial prefrontal cortex (RMPFC)	68
4.8	EEG electrodes designation for a 64-channels setup	69
4.9	HS risky over HS participant ERP; P300 peak: R:3µV, P:1.5µV / FZ Channel	70
4.10	LS participant Vs. LS risky ERP; P300 peak: P:42µV, R:9µV / FCZ Channel	72
4.11	HS risky Vs. LS risky ERP; P300 peak: HS R: 7.5µV, LS R:3.5µV / CP1 Channel	73
4.12	CNV region of HS vs. LS in Participant condition	74
4.13	CNV region of HS vs. LS in Risky condition	75
4.14	Correlation of SMS-M scores over P-300 peak values during "Risky" cue	77
4.15	Correlation of SMS-M scores over CNV peak values during "Risky" cu	77

LIST OF APPENDICES

Appendix Pa			Page	
	A	Ethical approval from the Medical Research Ethics committee for Human Research, Faculty Medicine and Health Sciences,		
		Universiti Futra Malaysia	101	
	В	Proforma Document	102	
	С	Smartphone addiction Scale-Malay version questionnaire	103	
	D	Modified Instagram Addiction Scale (IAS)	104	
	E	Research Grant	105	
	F	Pictures in the Instagram 'Risky' cues for fMRI study		
		thesis)	106	
	G	Patient Information Sheet for fMRI	107	
	Н	Consent by Patient for Clinical Research	111	

6

LIST OF ABBREVIATIONS

AC	Anterior Commissure
BOLD	Blood Oxygen Level Dependent Imaging
CFA	Confirmatory Factor Analyses
CNV	Contingent Negative Variation
DCM	Dynamic Causal Modelling
Deoxy-Hb	Deoxy hemoglobin
DLPFC	Dorso-lateral Prefrontal Cortex
DRARSF	Data Processing Assistant for Resting-State fMRI
EEG	Electroencephalography
EFA	Exploratory Factor Analyses
EP	Evoked Potentials
ERP	Event Related Potentials
E-SAPS	Estonian Smartphone Addiction Proneness Scale
FFT	Fast-Fourier Transform
fMRI	Functional Magnetic resonance Imaging
GLM	General Linear Model
НС	Healthy Cases
HRF	Haemodynamic Response Function
IAS	Instagram Addiction Scale
КМО	Kaiser-Meyer-Olkin
LIOG	Left Inferior Occipital Gyrus
LLG	Left Lingual Gyrus
LMOG	Left Middle Occipital Gyrus

LORETA	Low-resolution Brain Electromagnetic Tomography
LS	Low Scorer
MPFC	Medial Prefrontal Cortex
OCD	Obsessive Compulsive Disorder
OFC	Orbito Frontal Cortex
Oxy-Hb	Oxy hemoglobin
PC	Posterior Commissure
PET	Positron Emission Tomography
PFC	Prefrontal Cortex
PIGU	Problematic Instagram Usage
PPDN	Centre for Diagnostic Nuclear Imaging
PSC	Percentage Signal change
RC	Right Cuneus
Rc	Right Cerebellum
RFG	Right Fusiform Gyrus
RITG	Right Inferior Temporal Gyrus
ROI	Region of Interest
SAS	Smartphone Addiction Scale
SAS-M	Smartphone Addiction Scale- Malay Version
SNAs	Social Networking Applications
SPA	Smartphone Addiction
tb-fMRI	Task-based fMRI
TE	Time to Echo
TR	Time to Repeat
UPM	Universiti Putra Malaysia

CHAPTER 1

INTRODUCTION

1.1 Background of Study

Nowadays, when we talk about modern technology, we are referring to the application of scientific knowledge for practical purposes to help us in our daily lives. Technology is vital because it is used for nearly everything. The fact that human beings cannot stay apart from technologies such as computer systems, televisions, mobile phones cannot be denied anymore. The usage of technological innovations and devices have slowly taken an indispensable role in people's everyday lives; hence being without them would be unimaginable to some of us. Invention of mobile technology is one of those technology that has brought us closer than ever. Smartphones have evolved and developed swiftly over the past few years. Indeed a cell phone device has actually turned from being a simple pager to a smartphone device that provides tools such as GPS navigation, web browser, communication by instant messaging and also perform as a handheld gaming console (Patil & Karhe ,November 2012).

Based on statistics, in year 2016, an expected 62.9% of the population worldwide already owned at least one smartphone device and this figure is estimated to pass the five billion mark by the end of 2019 (Samaha & Hawi, 2016). It can be seen that smartphones have changed the manner in which we cooperate with one another significantly. Those changes first arose on social media networks and that is why social applications or apps are so important these days. Social apps like Facebook, Instagram and WhatsApp with 2.5 billion, 1.5 billion and 1 billion respectively, have the highest number of active users among all social apps (Salinas, 2018, August 08). Despite of the advantages to our daily life.

In fact, mobile dependency or smartphone addiction (SPA) is a kind of dependence syndrome that is observed among smartphone users in the younger generation. It can be seen as a spectrum of problematic smartphone usage with SPA being in the far end of the spectrum. According to Gazelle survey, "More than 25% of respondents reported that they 'almost always' use their smartphone while in a social setting such as during a meal or during a party and 58% said they use it 'usually' during these settings" (Belardi, 15 October 2013). Another study took place at the Alabama State University regarding the consequences of smartphone usage on the students, which stated that "we are not addicted to smartphones themselves, but that we are addicted to the information, entertainment, and personal connections that a smartphone delivers" (Emanuel, 2015). Scientific evidence indicate that users often relate using smartphones with impaired concentration, reduced short term memory

retrieval, concentration, tiredness, dizziness, disturbance in sleep pattern and also headaches (Al-Khlaiwi & Meo, 2004). Exorbitant mobile phone usage could result in a chain reaction, influencing one factor of the individual's life and progresses to involve other aspects of their being, which begins with problems in social interactions that can then lead to stress and depression, and in the long run can cause a negative impact on habits of daily living that involve eating and sleeping right (Thomée, Härenstam, & Hagberg, 2011).

The past researches on smartphone addiction or problematic smartphone usage have emphasized on the factors contributing to smartphone addiction. In this study, the response to reward and the brain dopaminergic reward system assessment via task-based fMRI with simultaneous acquisition of EEG recordings in the fMRI scanner (EEG-fMRI) will be conducted to investigate the correlation of the two parameters with smartphone addiction among male UPM undergraduate students.

FMRI measures the brain activity in such way that we can identify the alterations, which are linked with blood flow and oxygen supply in the brain. This technique is based on the certainty that whenever a region of the brain is being used, brain blood flow and neuronal activation are combined and blood stream to that particular region will increase (Thomée et al., 2011). Basically, the functional MRI utilizes the blood-oxygen-level dependent (BOLD) imaging inherent contrast that was initially discovered by the research team headed by Ogawa (Nikos K. Logothetis, Pauls, Augath, Trinath, & Oeltermann, 2001). fMRI scanning will map out activity of the neural in the cerebrum and / or the spinal cord by the means of imaging blood flow changes; known as hemodynamic response factor (HRF) that corresponds to energy utilized by the neurons (Huettel, Song, McCarthy, G.(2009), & 2009). Functional MRI technique can be joined with other brain imaging procedures such as near-infrared spectroscopy (NIRS) to have perfected information about both oxyhemoglobin and deoxyhemoglobin or with electroencephalography (EEG) to measure the functional brain activity along with electrical activity of the cerebral neurons.

Additionally, electroencephalography (EEG) performs as a non-invasive medical test that records the neuronal electrical impulse activity, using small metallic discs known as electrodes which are positioned on the scalp. As a rule, the neuronal cells in the brain communicate through electrical impulses that are constantly active, notwithstanding the patient or the participant is in sleep mode. Henceforth, EEG estimates the changes in electrical voltage that are generated by ionic currents within the cerebral neuronal cells (Schomer, L., da Silva, & Lopes, 2011). In clinical terms, EEG ordinarily means storing of the brain's spontaneous electrical activity within a given time period (Schomer et al., 2011). EEG diagnostic applications center either on its spectral content or on the event related potentials (ERP). The EEG recording result is utilized to investigate the kind of neural motions or what we know as "brain waves" that are observed in the frequency domain of EEG signals. In recent years, fMRI with simultaneous

EEG have turned to be a perfectly strong techniques combination, in which both functional MRI and electroencephalography have complementary strengths, one with high spatial resolution and the another one with a very high temporal resolution (Huettel et al., 2009). Therefore, this study aims to explore the regions of neuronal activations and patterns of EEG waveforms during response to reward task among undergraduate students in Universiti Putra Malaysia, particularly male students with problematic Instagram usage using simultaneous EEG-fMRI imaging.

1.2 Problem Statement and Justification

Problematic smartphone usage has been linked with poor sleep quality, impaired daytime functioning as well as depressive signs (Lemola, Perkinson-Gloor, Brand, Dewald-Kaufmann, & Grob, 2015). Thus, it is alarming that the usage of smartphones among the young adults has become increasingly prevalent. As matter of fact, the prevalence of problematic smartphone usage among public university students in a medical faculty in Malaysia has been observed to be as high as 47% (Ching et al., 2015). In truth, the matter of problematic smartphone usage could pose a serious problem among the young adults of this generation predominantly due to excessive usage of social networking applications (SNAs), particularly SNAs such as Facebook or Instagram, which are widely available on smartphones (Kross et al., 2013). There is a neural evidence of dopaminergic response to reward disorder among other types of substance addiction but not much is known regarding the neural correlates of smartphone addiction. Therefore, this study will aim to detect the neuropsychological changes among undergraduate students in UPM who have problematic smartphone usage with regards to response to reward towards Instagram cues.

The assessment of problematic smartphone usage and digital addiction is conventionally conducted using validated questionnaires such as the Chen Internet Addiction Scale (CIAS), Smartphone Addiction Scale, Smartphone Addiction Scale – Malay version (SAS-M) and previously Instagram Addiction Scale (IAS) (Chen, Weng, Su, Wu, & Yang, 2003) (Ching et al., 2015).

With regards to the SAS-M questionnaire, the students with problematic smartphone usage can be identified with cutoff scores, whereby high scorers (HS) can be differentiated from low scorers (LS) for the identification of subjects with problematic smartphone usage. We can also assess SNA pathological usage, particularly related to pathological Instagram usage by utilizing the modified Instagram Addiction Scale (IAS) questionnaire (Lancy D'Souza, Samyukta A, & Tevin John Bivera, July 2018). This method can help to identify subjects with Problematic Instagram Usage (PIGU). Nevertheless, these methods sometimes have a limitation on objectively assessing problematic smartphone usage.

Combining functional MRI and EEG techniques is a novel initiative way that is able to objectively determine the activation of neurons and lay out patterns of the brain wave in different segments of it (C.-Y. Wee et al., 2014). Functional MRI utilizes blood oxygenation level dependent (BOLD) imaging technique that gives different signals in the brain based on oxygen utilization whether in resting-state or task-based activity that is seen in a task-based fMRI (tb-fMRI). It's worthwhile to mention that, addiction have impact on the modulation of dopaminergic activity in the meso-corticolimbic circuitry, particularly in the hypothalamus and amygdala, which can be objectively evaluated using fMRI (Goldstein et al., 2007).

Considering that addiction is caused by several domains namely development of dependence, tolerance, withdrawal, etc. we aimed to focus on the development of tolerance towards rewarding cues based on Instagram upload. Visual stimuli have been known to cause arousal based on the emotional valence elicited by the pictures (Gerdes et al., 2010). For instance images that are positive/ pleasant was correlated with emotional arousal and fMRI activations in the right caudate nucleus, nucleus accumbens and left dorsolateral prefrontal cortex and negative/ unpleasant images elicited stronger activations in the amygdala (Gerdes et al., 2010).

Furthermore, a study by Nitschke et al, 2004 identified mothers viewing rewarding pictures of their newborn had higher activations of their orbitofrontal cortex (Nitschke et al., 2004). The importance of predicting a reward for a group of individuals is based on a positive or negative value assigned by the group for a stimulus, whereby increased activity can be observed in the prefrontal cortex indicating stimulation of the dopaminergic reward system (Lena et al., 2005, Perogamvros & Schwartz, 2012) by utilizing a task that presents the subject with images of negative emotional valence or 'Risky' images, we believe we can identify differences of brain reward system activation in subjects with PIGU. Conversely, images with positive emotional valence or referred to as 'Participant' images are expected to elicit pleasant feelings and response in both PIGU and Control groups and should not overly stimulate the prefrontal cortex, orbitofrontal cortex and amygdala in the brain reward system.

Additionally, EEG patterns can defect brain waveform in response to addictive behaviors, particularly in response to a rewarding stimulus (Boecker-Schlier et al., 2017). EEG patterns are shown as peaks and troughs in a recording sheet with the amplitudes of fluctuations referred to as evoked response potential (ERP) i.e. it is the measurement of the electrophysiological responses particularly of the central nervous system to a myriad of stimuli. Specifically, we can measure the P300 response based on EEG readings, which is a peak that is elicited during decision-making and it represents an endogenous ERP that mirrors the participant's response to a given cue (van Dinteren, Arns, Jongsma, & Kessels, 2014). Contingent negativity variance (CNV) is another EEG parameter that is closely related to response to a given stimulus. We generally

refer to the time between an imperative stimulus and the behavioral response as the reaction time. With reference to EEG, the equivalent to response time is the CNV, which is referred as thus due to the timing of the variation of the negatively deflected wave that depicts the elicited response, is contingent on the statistical relationship between the cue/ warning to stimulate anticipation and the imperative presentation of stimuli to assess response to reward. (Walter, Cooper, Aldridge, McCallum, & Winter, 1964).

Hence, combining these two techniques, fMRI BOLD-imaging with simultaneous-EEG can potentially to be an alternative and non-invasive biological marker for the evaluation of problematic smartphone usage.

1.3 Objectives

General Objective: The key objective is to correlate fMRI activation with EEG data, in response to cue related reactivity during response to reward among PIGU group.

Sub-Objectives:

- i. To identify the subjects' demographic data (age, gender, frequency of smartphone usage, SAS-M scores, modified IAS scores)
- ii. To identify percentage signal change (PSC) in regions of interest in tbfMRI study during response to reward (PIGU group vs Control group)
- iii. To compare the time of onset of CNV during response to reward (PIGU group vs Control group)
- iv. To compare the amplitude height of P300 peak wave during response to reward (PIGU group vs Control group)
- v. To correlate subject demographic data with EEG findings of P300 amplitude and PSC on tb-fMRI respectively
- vi. To correlate the EEG findings of P300 amplitude with PSC on tb-fMRI

1.4 Hypothesis

Hypothesis (H) and Null hypothesis (H0) for this thesis:

H1: There is a significant difference between the demographic data i.e., age, gender, and frequency of smartphone usage among the PIGU and controls.

H0: There is No significant difference between the demographic data i.e., age, gender, and frequency of smartphone usage among the PIGU and controls.

H2: There is a significant difference between the percentage signal change (PSC) in regions of interest in tb-fMRI study during response to reward among the PIGU group compared to the Control group.

H0: There is No significant difference between the percentage signal change (PSC) in regions of interest in tb-fMRI study during response to reward among the PIGU group compared to the Control group.

H3: There is a significant difference between the time of onset of CNV during response to reward among PIGU group compared to the Control group.

H0: There is No significant difference between the time of onset of CNV during response to reward among PIGU group compared to the Control group.

H4: There is a significant correlation between the subject demographic data with EEG findings of P300 amplitude and PSC on tb-fMRI in the PIGU group compared with the control group.

H0: There is No significant correlation between the subject demographic data with EEG findings of P300 amplitude and PSC on tb-fMRI in the PIGU group compared with the control group.

H5: There is a significant difference between the amplitude height of P300 peak wave during response to reward among the PIGU group compared to the Control group.

H0: There is No significant difference between the amplitude heights of P300 peak wave during response to reward among the PIGU group compared to the Control group.

H6: There is a significant correlation of the EEG findings of P300 amplitude with PSC on tb-fMRI during viewing 'Risky' cues in the PIGU group.

H0: There is No significant correlation of the EEG findings of P300 amplitude with PSC on tb-fMRI during viewing 'Risky' cues in the PIGU group.

1.5 Organization of Thesis

The outline of this thesis is as follows:

The first chapter explains the concepts for this response to reward study and Figure 1.1, gives the conceptual framework of this study.

In the following chapter, we provide the literature review from precedent works in studying types of assessments for smartphone addiction and Instagram addiction as well as fMRI and EEG studies related substance addiction, behavioral addictions and problematic smartphone usage. We will provide sufficient explanation from previous studies in order to answer, why this study focuses mainly on undergraduate male students in Malaysia and in particular to Instagram application. We will also discuss the significant EEG and fMRI findings from various addiction studies and also discuss the significance of using tb-fMRI in addiction studies. We will also debate the significance of using simultaneous EEG-fMRI in this study. We will highlight the relationship between problematic smartphone usage particularly PIGU, among male undergraduate students, with fMRI percentage signal change (PSC) in areas of the cerebrum involved in responding to reward stimuli and correlate this with simultaneous-EEG detected P300 wave and CNV.

Chapter 3 brings a concise explanation on the methods employed to design our study protocol. We will clarify our study materials and method used. We will underscore the procedures associated with tb-fMRI along with simultaneous-EEG data analysis and describe the statistical methods that we utilized in analyzing the study results.

In Chapter 4, we present our results from simultaneous EEG-fMRI study in the form of tables, graphs, MRI image figures as well as EEG signal recordings.

In Chapter 5, we emphasize on the discussion related to the simultaneous EEGfMRI study that was conducted among male undergraduate students in Universiti Putra Malaysia in relation to PIGU. We also contrast our results based on the simultaneous EEG-fMRI findings from other types of addiction studies and activation findings based on other task-based studies pertaining to addictions and response to reward. In this chapter, we will also give a synopsis of the thesis as well as expound on the important observations made in this study and the path for future works. In Chapter 6, we will summarize this thesis and conclude our limitations and the recommendations for future works. A brief biodata of student is provided at the end.



1.6 Conceptual Framework



REFERENCES

- Al-Khlaiwi, T., & Meo, S. A. (2004). Association of mobile phone radiation with fatigue, headache, dizziness, tension and sleep disturbance in Saudi population. Saudi Med J, 25(6), 732-736. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/15195201.
- Alpay, G., Goerke, M., & Sturmer, B. (2009). Precueing imminent conflict does not override sequence-dependent interference adaptation. Psychol Res, 73(6), 803-816. doi:10.1007/s00426-008-0196-9.
- Association, A. P. (2013). Diagnostic and Statistical Manual Mental Disorders,5th Edition. Retrieved from https://dsm.psychiatryonline.org.
- Adams, S. K., & Kisler, T. S. (2013). Sleep quality as a mediator between technology-related sleep quality, depression, and anxiety. Cyberpsychology, behavior and social networking, 16(1), 25-30. doi:10.1089/cyber.2012.0157.
- Azlina Wati Nikmat, Nurul Azreen Hashim, Muhammad Farid Saidi , Nur Suhailah Mohd Zaki, Nur Nabihah Hasan Shukri , & Abdulla, N. B. (2018). THE USE AND ADDICTION TO SMART PHONES AMONG MEDICAL STUDENTS AND STAFFS IN A PUBLIC UNIVERSITY IN MALAYSIA. ASEAN Journal of Psychiatry,, 19 (1), 8.
- Alosaimi, F. D., Alyahya, H., Alshahwan, H., Al Mahyijari, N., & Shaik, S. A. (2016). Smartphone addiction among university students in Riyadh, Saudi Arabia. Saudi medical journal, 37(6), 675-683. doi:10.15537/Smj.2016.6.14430.
- Babadi-Akashe, Z., Zamani, B. E., Abedini, Y., Akbari, H., & Hedayati, N. (2014). The Relationship between Mental Health and Addiction to Mobile Phones among University Students of Shahrekord, Iran. Addict Health, 6(3-4), 93-99.
- Balodis, I. M., Kober, H., Worhunsky, P. D., Stevens, M. C., Pearlson, G. D., & Potenza, M. N. (2012). Diminished frontostriatal activity during processing of monetary rewards and losses in pathological gambling. Biol Psychiatry, 71(8), 749-757. doi:10.1016/j.biopsych.2012.01.006.
- Becker, A., Kirsch, M., Gerchen, M. F., Kiefer, F., & Kirsch, P. (2017). Striatal activation and frontostriatal connectivity during non-drug reward anticipation in alcohol dependence. Addict Biol, 22(3), 833-843. doi:10.1111/adb.12352.
- Belardi, B. E. (15 October 2013). Consumers Crave iPhone More than facebook and Sex. Retrieved from https://www.prnewswire.com/newsreleases/consumers-crave-iphone-more-than-facebook-sex-accordingto-gazelle-159430685.html.

- Blackwood, D. H., & Muir, W. J. (1990). Cognitive brain potentials and their application. Br J Psychiatry Suppl(9), 96-101.
- Boecker-Schlier, R., Holz, N. E., Hohm, E., Zohsel, K., Blomeyer, D., Buchmann, A. F., Laucht, M. (2017). Association between pubertal stage at first drink and neural reward processing in early adulthood. Addict Biol, 22(5), 1402-1415. doi:10.1111/adb.12413.
- Cha, S.-S., & Seo, B.-K. (2018). Smartphone use and smartphone addiction in middle school students in Korea: Prevalence, social networking service, and game use. Health psychology open, 5(1), 2055102918755046. Retrieved from http://europepmc.org/abstract/MED/29435355.
- Ceballos, N. A., Bauer, L. O., & Houston, R. J. (2009). Recent EEG and ERP findings in substance abusers. Clinical EEG and neuroscience, 40(2), 122-128. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/19534304https://www.ncbi.nlm.nih.gov/pmc/PMC2746385/. doi:10.1177/155005940904000210.
- Cheever, N. A., Rosen, L. D., Carrier, L. M., & Chavez, A. (2014). Out of sight is not out of mind: The impact of restricting wireless mobile device use on anxiety levels among low, moderate and high users. Computers in Human Behavior, 37, 290-297. doi:10.1016/j.chb.2014.05.002.
- Courtney, K. E., Ghahremani, D. G., London, E. D., & Ray, L. A. (2014). The association between cue-reactivity in the precuneus and level of dependence on nicotine and alcohol. Drug Alcohol Depend, 141, 21-26. doi:10.1016/j.drugalcdep.2014.04.026.
- Chen, C. Y., Yen, J. Y., Wang, P. W., Liu, G. C., Yen, C. F., & Ko, C. H. (2016). Altered Functional Connectivity of the Insula and Nucleus Accumbens in Internet Gaming Disorder: A Resting State fMRI Study. Eur Addict Res, 22(4), 192-200. doi:10.1159/000440716.
- Chen, S.-H., Weng, L.-J., Su, Y.-J., Wu, H.-M., & Yang, P.-F. (2003). Development of a Chinese Internet Addiction Scale and Its Psychometric Study. Chinese Journal of Psychology, 45(3), 279-294.
- Ching, S. M., Yee, A., Ramachandran, V., Sazlly Lim, S. M., Wan Sulaiman, W. A., Foo, Y. L., & Hoo, F. K. (2015). Validation of a Malay Version of the Smartphone Addiction Scale among Medical Students in Malaysia. PLoS One, 10(10), e0139337. doi:10.1371/journal.pone.0139337.
- Chou, I. h. (2008). Read my mind. Nature Physics, 4(S1), S17-S17. doi:10.1038/nphys874.
- Colaner, & Seth. (August, 2012). Your Tablet and Smartphone Could Be Ruining Your Sleep. Retrieved from https://hothardware.com/news/your-tabletand-smartphone-are-ruining-your-sleep-.

- Davey, S., & Davey, A. (2014). Assessment of Smartphone Addiction in Indian Adolescents: A Mixed Method Study by Systematic-review and Metaanalysis Approach. Int J Prev Med, 5(12), 1500-1511.
- Dawson, G. D. (1947). Investigations on a patient subject to myoclonic seizures after sensory stimulation. Journal of neurology, neurosurgery, and psychiatry, 10(4), 141-162. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/18905644
- De Ridder, D., Manning, P., Leong, S. L., Ross, S., Sutherland, W., Horwath, C., & Vanneste, S. (2016). The brain, obesity and addiction: an EEG neuroimaging study. Sci Rep, 6, 34122. doi:10.1038/srep34122.
- De Ridder, D., Vanneste, S., Kovacs, S., Sunaert, S., & Dom, G. (2011). Transient alcohol craving suppression by rTMS of dorsal anterior cingulate: an fMRI and LORETA EEG study. Neurosci Lett, 496(1), 5-10. doi:10.1016/j.neulet.2011.03.074.
- Deco, G., Rolls, E., & Horwitz, B. (2004). "What" and "Where" in Visual Working Memory: A Computational Neurodynamical Perspective for Integrating fMRI and Single-Neuron Data (Vol. 16).
- Desmond, J. E., & Glover, G. H. (2002). Estimating sample size in functional MRI (fMRI) neuroimaging studies: statistical power analyses. J Neurosci Methods, 118(2), 115-128.
- Dieter, J., Hoffmann, S., Mier, D., Reinhard, I., Beutel, M., Vollstadt-Klein, S., Lemenager, T. (2017). The role of emotional inhibitory control in specific internet addiction - an fMRI study. Behav Brain Res, 324, 1-14. doi:10.1016/j.bbr.2017.01.046.
- Ding, W.-n., Sun, J.-h., Sun, Y.-w., Chen, X., Zhou, Y., Zhuang, Z.-g., Du, Y.-s. (2014). Trait impulsivity and impaired prefrontal impulse inhibition function in adolescents with internet gaming addiction revealed by a Go/No-Go fMRI study. Behavioral and Brain Functions, 10(1), 20. Retrieved from https://doi.org/10.1186/1744-9081-10-20. doi:10.1186/1744-9081-10-20.
- Dong, G., Devito, E. E., Du, X., & Cui, Z. (2012). Impaired inhibitory control in 'internet addiction disorder': a functional magnetic resonance imaging study. Psychiatry Res, 203(2-3), 153-158. doi:10.1016/j.pscychresns.2012.02.001.
- Dong, G., Zhou, H., & Zhao, X. (2010). Impulse inhibition in people with Internet addiction disorder: electrophysiological evidence from a Go/NoGo study. Neurosci Lett, 485(2), 138-142. doi:10.1016/j.neulet.2010.09.002.

- De-Sola, J., Talledo, H., Rodríguez de Fonseca, F., & Rubio, G. (2017). Prevalence of problematic cell phone use in an adult population in Spain as assessed by the Mobile Phone Problem Use Scale (MPPUS). PLoS One, 12(8), e0181184. Retrieved from https://doi.org/10.1371/journal.pone.0181184. doi:10.1371/journal.pone.0181184.
- E.Raichle, & Marcus. (9 August 2013). Brain regions active when our minds wander may hold a key to understanding neurological disorders and even consciousness itself. Retrieved from https://www.scientificamerican.com/article/the-brains-dark-energy/
- Emanuel, R. (2015). The truth about smartphone addiction (Vol. 49).
- Ferrazzoli, D., Albanese, M., Sica, F., Romigi, A., Sancesario, G., Marciani, M. G., Placidi, F. (2013). Electroencephalography and dementia: a literature review and future perspectives. CNS Neurol Disord Drug Targets, 12(4), 512-519.
- Giroldini, W., Pederzoli, L., Bilucaglia, M., Caini, P., Ferrini, A., Melloni, S., Tressoldi, P. (2016). EEG correlates of social interaction at distance. F1000Research, 4, 457-457. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/26966513
- Goldstein, R. Z., Tomasi, D., Rajaram, S., Cottone, L. A., Zhang, L., Maloney, T., Volkow, N. D. (2007). Role of the anterior cingulate and medial orbitofrontal cortex in processing drug cues in cocaine addiction. Neuroscience, 144(4), 1153-1159. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/17197102 https://www.ncbi.nlm.nih.gov/pmc/PMC1852512/.doi:10.1016/j.neuroscie nce.2006.11.024.
- Grin-Yatsenko, V. A., Baas, I., Ponomarev, V. A., & Kropotov, J. D. (2010). Independent component approach to the analysis of EEG recordings at early stages of depressive disorders. Clin Neurophysiol, 121(3), 281-289. doi:10.1016/j.clinph.2009.11.015.
- Gerdes, A., Wieser, M., Muehlberger, A., Weyers, P., Alpers, G., Plichta, M., Pauli, P. (2010). Brain Activations to Emotional Pictures are Differentially Associated with Valence and Arousal Ratings. Frontiers in Human Neuroscience, 4(175). Retrieved from https://www.frontiersin.org/article/10.3389/fnhum.2010.00175. doi:10.3389/fnhum.2010.00175.
- Gnambs, T., & Appel, M. (2018). Narcissism and Social Networking Behavior: A Meta-Analysis. J Pers, 86(2), 200-212. doi:10.1111/jopy.12305.
- Gordon, H. W. (2016). Laterality of Brain Activation for Risk Factors of Addiction. Current drug abuse reviews, 9(1), 1-18. Retrieved from http://europepmc.org/abstract/MED/26674074.

- Griffiths, M., & Kuss, D. (2017). Adolescent social media addiction (revisited) (Vol. 35).
- Hanlon, C. A., & Canterberry, M. (2012). The use of brain imaging to elucidate neural circuit changes in cocaine addiction. Substance abuse and rehabilitation, 3(1), 115-128. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/23162375

https://www.ncbi.nlm.nih.gov/pmc/PMC3499034/. doi:10.2147/SAR.S35153. Hatton, & Celia. (September 2014). Chongqing's 'mobile lane'. Retrieved from https://www.bbc.com/news/blogs-china-blog-29201934.

- Haber, S. N., & Calzavara, R. (2009). The cortico-basal ganglia integrative network: the role of the thalamus. Brain Res Bull, 78(2-3), 69-74.doi:10.1016/j.brainresbull.2008.09.013.
- Hookham, Mark, Togoh, Isabel, Yeates, & Alex. (February 2016). Walkers hit by curse of the smomble. Retrieved from https://www.thetimes.co.uk/article/walkers-hit-by-curse-of-the-smomble-kjpxbhm6cb0.
- Ha, Y. W., Kim, J., Libaque-Saenz, C. F., Chang, Y., & Park, M.-C. (2015). Use and gratifications of mobile SNSs: Facebook and KakaoTalk in Korea. Telematics and Informatics, 32(3), 425-438. Retrieved from http://www.sciencedirect.com/science/article/pii/S0736585314000744. doi:https://doi.org/10.1016/j.tele.2014.10.006.
- Huettel, S. A., Song, A. W., McCarthy, G.(2009), & (2009). Functional Magnetic Resonance.
- Huang Anna, S., Mitchell Jameson, A., Haber Suzanne, N., Alia-Klein, N., & Goldstein Rita, Z. (2018). The thalamus in drug addiction: from rodents to humans. Philosophical Transactions of the Royal Society B: Biological Sciences, 373(1742), 20170028. doi:10.1098/rstb.2017.0028.
- Haug, S., Castro, R. P., Kwon, M., Filler, A., Kowatsch, T., & Schaub, M. P. (2015). Smartphone use and smartphone addiction among young people in Switzerland. J Behav Addict, 4(4), 299-307. doi:10.1556/2006.4.2015.037.
- Hanlon, C., Dowdle, L., Naselaris, T., Canterberry, M., & Cortese, B. (2014). Hanlon CA, Dowdle LT, Naselaris T, Canterberry M, Cortes BM. Visual cortex activation to drug cues: a meta-analysis of functional neuroimaging papers in addiction and substance abuse literature. Drug Alcohol Depend 143: 206-212 (Vol. 143).

- Hayashi, T., Ko, J. H., Strafella, A. P., & Dagher, A. (2013). Dorsolateral prefrontal and orbitofrontal cortex interactions during self-control of cigarette craving. Proceedings of the National Academy of Sciences, 110(11), 4422. Retrieved from http://www.pnas.org/content/110/11/4422.abstract. doi:10.1073/pnas.1212185110.
- Https://www.ncbi.nlm.nih.gov/pmc/PMC4028909/.doi:10.4103/0971-3026.130686.
- Https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3188411/.doi:10.1016/j.neuron. 2011.01.027.
- Https://www.ncbi.nlm.nih.gov/pmc/articles/PMC2242347/.doi:10.1111/j.1469-8986.2006.00376.x.
- Https://www.ncbi.nlm.nih.gov/pmc/PMC4770988/.doi:10.12688/f1000research. 6755.3.
- Https://doi.org/10.1089/cyber.2012.0260.doi:10.1089/cyber.2012.0260.
- Https://www.ncbi.nlm.nih.gov/pmc/articles/PMC4770183/.doi:10.4329/wjr.v8.i2. 210.
- Imaging. The Yale Journal of Biology and Medicine, 82. Retrieved from https://www.ncbi.nlm.nih.gov/pmc/articles/PMC2794501/.
- Ibañez, D. (November,2015). What is Loreta? What can I use it for? . Retrieved from https://www.neuroelectrics.com/blog/what-is-loreta-what-can-i-use-it-for/.
- Ithnain, N., Ghazali, S., & Jaafar, N. (2018). Relationship between Smartphone Addiction with Anxiety and Depression among Undergraduate Students in Malaysia (Vol. 8).
- James, J. S., Rajesh, P., Chandran, A. V., & Kesavadas, C. (2014). fMRI paradigm designing and post-processing tools. The Indian journal of radiology & imaging, 24(1), 13-21. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/24851001.

Jiménez-Arriero;, M. A., Rodríguez-Torresano;, J., Ponce;, G., Hoenicka;, J., Rodríguez;, R., Rubio;, G., & Palomo, T. (2009). P300 in alcohol dependence: Effects of TaqI-A genotype. The European Journal of Psychiatry. Retrieved from http://scielo.isciii.es/scielo.php?script=sci_arttext&pid=S0213-61632009000400003.

- Jo, H.-s., Na, E., & Kim, D.-J. (2018). The relationship between smartphone addiction predisposition and impulsivity among Korean smartphone users. Addiction Research & Theory, 26(1), 77-84. Retrieved from https://doi.org/10.1080/16066359.2017.1312356.doi:10.1080/16066359. 2017.1312356.
- Karch, S., Loy, F., Krause, D., Schwarz, S., Kiesewetter, J., Segmiller, F., Pogarell, O. (2016). Increased Event-Related Potentials and Alpha-, Beta-, and Gamma-Activity Associated with Intentional Actions. Frontiers in Psychology, 7(7). Retrieved from https://www.frontiersin.org/article/10.3389/fpsyg.2016.00007. doi:10.3389/fpsyg.2016.00007.
- Mak, K. K., Lai, C. M., Ko, C. H., Chou, C., Kim, D. I., Watanabe, H., & Ho, R. C. (2014). Psychometric properties of the Revised Chen Internet Addiction Scale (CIAS-R) in Chinese adolescents. J Abnorm Child Psychol, 42(7), 1237-1245. doi:10.1007/s10802-014-9851-3.
- Nisha N, Ibrahim B, Sharifat H, Abdul Rashid A, Suppiah S. (2019) Incremental benefits of EEG informed fMRI in the study of disorders related to mesocorticolimbic dopamine pathway dysfunction: A systematic review of recent literature. J Clin Neurosci.;65:87-99.
- Keilholz, S. D., Magnuson, M., & Thompson, G. (2010). Evaluation of data-driven network analysis approaches for functional connectivity MRI. Brain Struct Funct, 215(2), 129-140. doi:10.1007/s00429-010-0276-7.
- Kononowicz, T. W., & Penney, T. B. (2016). The contingent negative variation (CNV): timing isn't everything. Current Opinion in Behavioral Sciences, 8, 231-237. Retrieved from http://www.sciencedirect.com/science/article/pii/S2352154616300456. doi:https://doi.org/10.1016/j.cobeha.2016.02.022.
- Koob, G. F., & Volkow, N. D. (2010). Neurocircuitry of addiction. Neuropsychopharmacology, 35(1), 217-238. doi:10.1038/npp.2009.110.
- Ko, C.-H., Yen, J.-Y., Yen, C.-F., Lin, H.-C., & Yang, M.-J. (2007). Factors Predictive for Incidence and Remission of Internet Addiction in Young Adolescents: A Prospective Study (Vol. 10).
- Kross, E., Verduyn, P., Demiralp, E., Park, J., Lee, D. S., Lin, N., Ybarra, O. (2013). Facebook Use Predicts Declines in Subjective Well-Being in Young Adults. PLoS One, 8(8), e69841. Retrieved from https://doi.org/10.1371/journal.pone.0069841. doi:10.1371/journal.pone.0069841.

KircaburunKagan, & D., G. (2018). Instagram addiction and the Big Five of personality: The mediating role of self-liking. J Behav Addict, 7(1), 158-170. Retrieved from https://akademiai.com/doi/abs/10.1556/2006.7.2018.15. doi:10.1556/2006.7.2018.15.

Kuss, D., & Griffiths, M. (2012). Internet and Gaming Addiction: A Systematic Literature Review of Neuroimaging Studies (Vol. 2).

Kwon, M., Kim, D.-J., Cho, H., & Yang, S. (2014). The Smartphone Addiction Scale: Development and Validation of a Short Version for Adolescents. PLoS One, 8(12), e83558. Retrieved from https://doi.org/10.1371/journal.pone.0083558. doi:10.1371/journal.pone.0083558.

- Kwon, M., Kim, D.-J., Cho, H., & Yang, S. (2013). The smartphone addiction scale: development and validation of a short version for adolescents. PLoS One, 8(12), e83558-e83558. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/24391787.
- Lehmann, D., & Skrandies, W. (1984). Spatial analysis of evoked potentials in man--a review. Prog Neurobiol, 23(3), 227-250.
- Light, G. A., Williams, L. E., Minow, F., Sprock, J., Rissling, A., Sharp, R., Braff, D. L. (2010). Electroencephalography (EEG) and event-related potentials (ERPs) with human participants. Curr Protoc Neurosci, Chapter 6, Unit 6.25.21-24. doi:10.1002/0471142301.ns0625s52.
- Lemola, S., Perkinson-Gloor, N., Brand, S., Dewald-Kaufmann, J. F., & Grob, A. (2015). Adolescents' electronic media use at night, sleep disturbance, and depressive symptoms in the smartphone age. J Youth Adolesc, 44(2), 405-418. doi:10.1007/s10964-014-0176.
- Logothetis, N. K. (2008). What we can do and what we cannot do with fMRI. Nature, 453(7197), 869-878. doi:10.1038/nature06976.
- Lena, I., Parrot, S., Deschaux, O., Muffat-Joly, S., Sauvinet, V., Renaud, B., Gottesmann, C. (2005). Variations in extracellular levels of dopamine, noradrenaline, glutamate, and aspartate across the sleep--wake cycle in the medial prefrontal cortex and nucleus accumbens of freely moving rats. J Neurosci Res, 81(6), 891-899. doi:10.1002/jnr.20602.
- Logothetis, N. K., Pauls, J., Augath, M., Trinath, T., & Oeltermann, A. (2001). Neurophysiological investigation of the basis of the fMRI signal. Nature, 412(6843), 150-157. Retrieved from https://doi.org/10.1038/35084005. doi:10.1038/35084005.
- Luck, S. J. (August 2005). An Introduction to the Event-Related Potential by: Steven A. HillyardDepartment of Neurosciences, University of California, San Diego.

- Lopez-Fernandez, O., Honrubia-Serrano, L., Freixa-Blanxart, M., & Gibson, W. (2014). Prevalence of problematic mobile phone use in British adolescents. Cyberpsychology, behavior and social networking, 17(2), 91-98. Retrieved from http://europepmc.org/abstract/MED/23981147.
- Liu, J., Li, W., Zhou, S., Zhang, L., Wang, Z., Zhang, Y., Li, L. (2016). Functional characteristics of the brain in college students with internet gaming disorder. Brain Imaging Behav, 10(1), 60-67. doi:10.1007/s11682-015-9364-x.
- Lancy D'Souza, Samyukta A, & Tevin John Bivera. (July 2018). Development and Validation of Test for Instagram Addiction (TIA). The international journal of Indian psychology. doi: 10.25215/0603.81.
- Majeed, W., Magnuson, M., & Keilholz, S. D. (2009). Spatiotemporal dynamics of low frequency fluctuations in BOLD fMRI of the rat. J Magn Reson Imaging, 30(2), 384-393. doi:10.1002/jmri.21848.
- Matousek, M. (1991). EEG patterns in various subgroups of endogenous depression. Int J Psychophysiol, 10(3), 239-243.
- Matar Boumosleh, J., & Jaalouk, D. (2017). Depression, anxiety, and smartphone addiction in university students- A cross sectional study. PLoS One, 12(8), e0182239. Retrieved from https://doi.org/10.1371/journal.pone.0182239. doi:10.1371/journal.pone.0182239.
- Mento, G. (2013). The passive CNV: carving out the contribution of task-related processes to expectancy. Frontiers in Human Neuroscience, 7(827). Retrieved

from:https://www.frontiersin.org/article/10.3389/fnhum.2013.00827.doi:10 .3389/fnhum.2013.00827.

Noack, R. (April, 2016). This city embedded traffic lights in the sidewalks so that smartphone users don't have to look up The Washington Post. Retrieved from

https://www.washingtonpost.com/news/worldviews/wp/2016/04/25/thiscity-embedded-traffic-lights-in-the-sidewalks-so-that-smartphone-usersdont-have-to-look-up/?utm_term=.91120eae8a9f.

- Nitschke, J. B., Nelson, E. E., Rusch, B. D., Fox, A. S., Oakes, T. R., & Davidson,
 R. J. (2004). Orbitofrontal cortex tracks positive mood in mothers viewing pictures of their newborn infants. Neuroimage, 21(2), 583-592. doi:10.1016/j.neuroimage.2003.10.005.
- Organization, W. H. (December, 2010). Dependence syndrome. Retrieved from https://www.who.int/substance_abuse/terminology/definition1/en/.

- Ogawa, S., Lee, T. M., Kay, A. R., & Tank, D. W. (1990). Brain magnetic resonance imaging with contrast dependent on blood oxygenation. Proceedings of the National Academy of Sciences of the United States of America, 87(24), 9868-9872. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/2124706.
- Patil, C. S., R. R. K., & , M. A. A. (November 2012). Development of Mobile Technology: A Survey. International Journal of Advanced Research in Electrical, Electronics and Instrumentation Engineering, 1(5), 6.
- Pawlikowski, M., Altstötter-Gleich, C., & Brand, M. (2013). Validation and psychometric properties of a short version of Young's Internet Addiction Test. Computers in Human Behavior, 29(3), 1212-1223. doi:https://doi.org/10.1016/j.chb.2012.10.014.
- Peterson, R. L. (2005). The neuroscience of investing: fMRI of the reward system. Brain Research Bulletin, 67(5), 391-397. Retrieved from http://www.sciencedirect.com/science/article/pii/S036192300500225X. doi:https://doi.org/10.1016/j.brainresbull.2005.06.015.
- Perogamvros, L., & Schwartz, S. (2012). The roles of the reward system in sleep and dreaming. Neurosci Biobehav Rev, 36(8), 1934-1951. doi:10.1016/j.neubiorev.2012.05.010.
- Pierce, R. C., & Kumaresan, V. (2006). The mesolimbic dopamine system: the final common pathway for the reinforcing effect of drugs of abuse? Neurosci Biobehav Rev, 30(2), 215-238. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/16099045.doi:10.1016/j.neubiorev. 2005.04.016.
- Patrick, C. J., Bernat, E. M., Malone, S. M., Iacono, W. G., Krueger, R. F., & McGue, M. (2006). P300 amplitude as an indicator of externalizing in adolescent males. Psychophysiology, 43(1), 84-92. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/16629688.
- Picton, T. W. (1992). The P300 wave of the human event-related potential. J Clin Neurophysiol, 9(4), 456-479.
- Perlman, G., Markin, A., & Iacono, W. G. (2013). P300 amplitude reduction is associated with early-onset and late-onset pathological substance use in a prospectively studied cohort of 14-year-old adolescents. Psychophysiology, 50(10), 974-982. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/23905780.
- Roberts, J. A., Yaya, L. H., & Manolis, C. (2014). The invisible addiction: cellphone activities and addiction among male and female college students. J Behav Addict, 3(4), 254-265. doi:10.1556/jba.3.2014.015.

- Rushworth, Matthew F. S., Noonan, MaryAnn P., Boorman, Erie D., Walton, Mark E., & Behrens, Timothy E. (2011). Frontal Cortex and Reward-Guided Learning and Decision-Making. Neuron, 70(6), 1054-1069. Retrieved from http://www.sciencedirect.com/science/article/pii/S0896627311003953. doi:https://doi.org/10.1016/j.neuron.2011.05.014.
- Salinas, S. (2018, August 08). PEAK SOCIAL? The major social platforms are showing a significant slowdown in users. Retrieved from https://www.cnbc.com/2018/08/08/social-media-active-users-around-the-world.html.
- Samaha, M., & Hawi, N. S. (2016). Relationships among smartphone addiction, stress, academic performance, and satisfaction with life. Computers in Human Behavior, 57, 321-325. doi:10.1016/j.chb.2015.12.045.
- Sanou, B. (2015). ICT Facts and Figures-The World in 2015. Retrieved from https://www.itu.int/en/ITUD/Statistics/Documents/facts/ICTFactsFigures2 015.pdf.
- Schomer, L., D., da Silva, & Lopes, F. (2011). Niedermeyer's Electroencephalography: Basic Principles, Clinical Applications, and Related Fields: Lippincott Williams & Wilkins (LWW).
- Schultz, W. (2015). Neuronal Reward and Decision Signals: From Theories to Data. Physiol Rev, 95(3), 853-951. doi:10.1152/physrev.00023.2014.
- Shereena, V. B., & Raju, G. (2016, 23-25 March 2016). Literature review of fMRI image processing techniques. Paper presented at the 2016 International Conference on Wireless Communications, Signal Processing and Networking (WiSPNET).
- Sherman, L. E., Greenfield, P. M., Hernandez, L. M., & Dapretto, M. (2018). Peer Influence Via Instagram: Effects on Brain and Behavior in Adolescence and Young Adulthood. Child Dev, 89(1), 37-47. doi:10.1111/cdev.12838.
- Smeebsmeeb, AliceD.AliceD, MarkMark, & Amirsaman Sajad. (2015). Why is Fast Fourier Transform applied to raw EEG data? Retrieved from https://biology.stackexchange.com/questions/44955/why-is-fast-fourier transform-applied-to-raw-eeg-data.
- Sur, S., & Sinha, V. K. (2009). Event-related potential: An overview. Ind Psychiatry J, 18(1), 70-73. doi:10.4103/0972-6748.57865.
- Sepede, G., Tavino, M., Santacroce, R., Fiori, F., Salerno, R. M., & Di Giannantonio, M. (2016). Functional magnetic resonance imaging of internet addiction in young adults. World journal of radiology, 8(2), 210-225. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/26981230

- Teixeira, T. (April 2010). Meet Marty Cooper the inventor of the mobile phone. Retrieved,from:http://news.bbc.co.uk/2/hi/programmes/click_online/8395 90.stm
- Thomée, S., Härenstam, A., & Hagberg, M. (2011). Mobile phone use and stress, sleep disturbances, and symptoms of depression among young adults a prospective cohort study. BMC Public Health, 11(1), 66. Retrieved from https://doi.org/10.1186/1471-2458-11-66. doi:10.1186/1471-2458-11-66.
- Tang, C. S., Koh, Y. W., & Gan, Y. (2017). Addiction to Internet Use, Online Gaming, and Online Social Networking Among Young Adults in China, Singapore, and the United States. Asia Pac J Public Health, 29(8), 673-682. doi:10.1177/1010539517739558.
- Tran, T. D., Tran, T., & Fisher, J. (2013). Validation of the depression anxiety stress scales (DASS) 21 as a screening instrument for depression and anxiety in a rural community-based cohort of northern Vietnamese women. BMC Psychiatry, 13, 24. doi:10.1186/1471-244x-13-24.
- UNSW, P. D. (3 January 2018). Depression, Anxiety and Stress Scale DASS 21. Retrieved from https://at-ease.dva.gov.au/veterans/resource-library/depression-anxiety-and-stress-scale-dass-21.
- Volkow, N. D., Koob, G. F., & McLellan, A. T. (2016). Neurobiologic Advances from the Brain Disease Model of Addiction. N Engl J Med, 374(4), 363-371. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/26816013. doi:10.1056/NEJMra1511480.
- Volkow, N. D., Baler, R. D., & Goldstein, R. Z. (2011). Addiction: pulling at the neural threads of social behaviors. Neuron, 69(4), 599-602. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/21338873.
- Van Dinteren, R., Arns, M., Jongsma, M. L. A., & Kessels, R. P. C. (2014). P300 development across the lifespan: a systematic review and meta-analysis. PLoS One, 9(2), e87347-e87347. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/24551055.
- Walter, W. G., Cooper, R., Aldridge, V. J., McCallum, W. C., & Winter, A. L. (1964). CONTINGENT NEGATIVE VARIATION: AN ELECTRIC SIGN OF SENSORIMOTOR ASSOCIATION AND EXPECTANCY IN THE HUMAN BRAIN. Nature, 203, 380-384. doi:10.1038/203380a0.
- Walsh, P., Kane, N., & Butler, S. (2005). The clinical role of evoked potentials. J Neurol Neurosurg Psychiatry, 76 Suppl 2, ii16-22. Retrieved from https://www.ncbi.nlm.nih.gov/pubmed/15961863. doi:10.1136/jnnp.2005.068130.

- Wang, L., Shen, H., Lei, Y., Zeng, L. L., Cao, F., Su, L., Hu, D. (2017). Altered default mode, fronto-parietal and salience networks in adolescents with Internet addiction. Addict Behav, 70, 1-6. doi:10.1016/j.addbeh.2017.01.021.
- Wee, C.-Y., Zhao, Z., Yap, P.-T., Wu, G., Shi, F., Price, T., Shen, D. (2014). Disrupted Brain Functional Network in Internet Addiction Disorder: A Resting-State Functional Magnetic Resonance Imaging Study. PLoS One, 9(9), e107306. Retrieved from https://doi.org/10.1371/journal.pone.0107306. doi:10.1371/journal.pone.0107306.
- Worhunsky, P. D., Potenza, M. N., & Rogers, R. D. (2017). Alterations in functional brain networks associated with loss-chasing in gambling disorder and cocaine-use disorder. Drug Alcohol Depend, 178, 363-371. doi:10.1016/j.drugalcdep.2017.05.025.
- Yager, L. M., Garcia, A. F., Wunsch, A. M., & Ferguson, S. M. (2015). The ins and outs of the striatum: role in drug addiction. Neuroscience, 301, 529-541. doi:10.1016/j.neuroscience.2015.06.033.
- Yang, Y. L., Deng, H. X., Xing, G. Y., Xia, X. L., & Li, H. F. (2015). Brain functional network connectivity based on a visual task: visual information processing-related brain regions are significantly activated in the task state. Neural Regen Res, 10(2), 298-307. doi:10.4103/1673-5374.152386.
- Yang, C.-c. (2016). Instagram Use, Loneliness, and Social Comparison Orientation: Interact and Browse on Social Media, But Don't Compare. Cyberpsychology, Behavior, and Social Networking, 19(12), 703-708. Retrieved from:https://www.liebertpub.com/doi/abs/10.1089/cyber.2016.0201.doi:1

0.1089/cyber.2016.0201.

Zhang, S., Li, X., Lv, J., Jiang, X., Ge, B., Guo, L., & Liu, T. (2015, 16-19 April 2015). Characterizing and differentiating task-based and resting state FMRI signals via two-stage dictionary learning. Paper presented at the 2015 IEEE 12th International Symposium on Biomedical Imaging (ISBI).

BIODATA OF STUDENT

Hamed Sharifat was born on March 13, 1993 in Iran. He received his high school certificate from Iranian Imam Khomeini School (Ampang, Malaysia / 2010-2012). He continued his Bachelor of Science in Medical Physics at USM University, Penang, Malaysia (2012-2016). After graduating, he enrolled as a full-time master's degree candidate in Faculty of Medicine & Health Sciences, UPM in Feb 2018.



LIST OF PUBLICATIONS

Journals

- Sharifat, H., Rashid, A. A., & Suppiah, S. (2018). Systematic Review of the Utility of Functional MRI to Investigate Internet Addiction Disorder: Recent Updates on Resting State and Task-Based fMRI. Malaysian Journal of Medicine and Health Sciences, 14(1),21–33. https://medic.upm.edu.my/upload/dokumen/2018031608380404_MJMH S_Vol14_No1_Jan2018_0080_-_2nd_proof.pdf
- Nasser, L.J., Rashid, H.S., Hamid, E.A., Andrew, S.M., & Suppiah, S. (2019). ASSESSMENT OF PROBLEMATIC FACEBOOK USE AMONG UNDERGRADUATE STUDENTS IN UPM CORRELATED WITH DEPRESSION, ANXIETY AND STRESS. INTERNATIONAL JOURNAL OF PUBLIC HEALTH AND CLINICAL SCIENCES, 6(4), 113–132. https://doi.org/10.32827/ijphcs.6.4.113
- Syed Nasser, N., Sharifat, H., Abdul Rashid, A., Mohamed, M., Ab Hamid, S., Abdul Rahim, E., Mustafa, S., Siew Mooi, C., Tyagi, R., Suppiah, S.(2019). SMARTPHONE ADDICTION AMONG UNDERGRADUATE STUDENTS IN UPM: A FUNCTIONAL MAGNETIC RESONANCE IMAGING STUDY. INSTITUTE FOR YOUTH RESEARCH MALAYSIA, 2, 151–171. http://www.ippbm.gov.my/images/2019/MJYS/Edisi%20Khas%20Agenda %20Sosial%20Vol%202.pdf
- Syed Nasser, N., Ibrahim, B., Sharifat, H., Abdul Rashid, A., & Suppiah, S. (2019). Incremental benefits of EEG informed fMRI in the study of disorders related to meso-corticolimbic dopamine pathway dysfunction: A systematic review of recent literature. Journal of clinical neuroscience : official journal of the Neurosurgical Society of Australasia, 65, 87–99. https://doi.org/10.1016/j.jocn.2019.03.054
- Nasser, N.S., Sharifat, H., Rashid, A., Hamid, S.A., Rahim, E.B., Mohamad, M., Tyagi, R., Ismail, S.I., Ching, S., & Suppiah, S. (2020). Validation of Emotional Stimuli Flashcards for Conducting 'Response to Reward' fMRI study among Malaysian students. medRxiv; DOI: 10.1101/2020.01.17.20017202.
- Nisha, S. N., Fathinul Fikri, A. S., Aida, A. R., Salasiah, M., Hamed, S., Rohit, T., Amei Farina, A. R., Loh, J. L., Mazlyfarina, M., & Subapriya, S. (2020). The objective assessment of the effects on cognition functioning among military personnel exposed to hypobaric-hypoxia: A pilot fMRI study. The Medical journal of Malaysia, 75(1), 62– 67.https://pubmed.ncbi.nlm.nih.gov/32008023/

- Nasser, N. S., Sharifat, H., Rashid, A. A., Hamid, S. A., Rahim, E. A., Loh, J. L., Ching, S. M., Hoo, F. K., Ismail, S., Tyagi, R., Mohammad, M., & Suppiah, S. (2020). Cue-Reactivity Among Young Adults With Problematic Instagram Use in Response to Instagram-Themed Risky Behavior Cues: A Pilot fMRI Study. Frontiers in psychology, 11, 556060. https://doi.org/10.3389/fpsyg.2020.556060
- Nasser, N. S., Loh, J. L., Rashid, A. A., Sharifat, H., Ahmad, U., Ibrahim, B., Mustafa, S., Hoo, F. K., Ching, S. M., & Suppiah S, S. (2020). A survey on smartphone dependence and psychological effects among undergraduate students in a Malaysian University. The Medical journal of Malaysia, 75(4), 356–362. http://www.e-mjm.org/2020/v75n4/smartphone-addiction.pdf

Conferences

Sharifat, H., Rashid, A. A., & Suppiah, S. (2019). Decoding the mystery ofdigital addiction: EEG-informed fMRI's role as a potential biomarker for response to reward assessment. Paper presented in Malaysian Congress of Radiology (MCOR2019) at Doubletree by Hilton, Kuala Lumpur, Malaysia 2019



UNIVERSITI PUTRA MALAYSIA

STATUS CONFIRMATION FOR THESIS / PROJECT REPORT AND COPYRIGHT

ACADEMIC SESSION : Second Semester 2020/2021

TITLE OF THESIS / PROJECT REPORT :

EXPLORING NEURONAL ACTIVATION IN RESPONSE TO REWARD THROUGH fMRI CORRELATED WITH SIMULTANEOUS EEG AMONG MALE UNDERGRADUATES HAVING PROBLEMATIC SMARTPHONE USAGE

NAME OF STUDENT: SHARIFAT HAMED

I acknowledge that the copyright and other intellectual property in the thesis/project report belonged to Universiti Putra Malaysia and I agree to allow this thesis/project report to be placed at the library under the following terms:

1. This thesis/project report is the property of Universiti Putra Malaysia.

- 2. The library of Universiti Putra Malaysia has the right to make copies for educational purposes only.
- 3. The library of Universiti Putra Malaysia is allowed to make copies of this thesis for academic exchange.

I declare that this thesis is classified as :

*Please tick (V)



CONFIDENTIAL



RESTRICTED



(Contain confidential information under Official Secret Act 1972).

(Contains restricted information as specified by the organization/institution where research was done).

I agree that my thesis/project report to be published as hard copy or online open access.

This thesis is submitted for :



Embargo from_____until _____ (date) (date)

Approved by:

(Signature of Student) New IC No/ Passport No.: (Signature of Chairman of Supervisory Committee) Name:

Date :

Date :

[Note : If the thesis is CONFIDENTIAL or RESTRICTED, please attach with the letter from the organization/institution with period and reasons for confidentially or restricted.]