INTRINSIC CONNECTIONS BETWEEN FRONTO-PARIETAL NETWORK DURING MENTAL ARITHMETIC TASKS

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UNIVERSITI TEKNOLOGI MALAYSIA

# INTRINSIC CONNECTIONS BETWEEN FRONTO-PARIETAL NETWORK DURING MENTAL ARITHMETIC TASKS 

## NOR SARADATUL AKMAR BINTI ZULKIFLI

A thesis submitted in fulfilment of the requirements for the award of the degree of Doctor of Philosophy

Specially dedicated to my beloved parents;
Zulkili bin Omar and Romlah Binti Abdullah, and my siblings
Who have encouraged me throughout my journey of education.
A warm thanks to all

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#### Abstract

Arithmetic knowledge are composed of several components, including procedural skill which are simple mathematics computation (MC) and complex mathematics computation, (MM), conceptual understanding and factual knowledge (FR). These components have been the basic references when conducting the neuroimaging study of cognitive arithmetic. Previous studies showed evidence of cortical activation of fronto-parietal regions with different intensity while performing various types of mental arithmetic problems. In addition, some studies have suggested that problem-size or difficulty can affect the activation of fronto-parietal regions. However, not much research has been done on effects of problem-size and arithmetic components in terms of their brain connectivity. Therefore this study was carried out to determine both empirical effects on the aspect of Functional Specialization (FS) and Effective Connectivity (EC) of fronto-parietal networks. In this study, functional Magnetic Resonance Imaging (fMRI) scans were performed on twenty-two ( $n=22$ ) healthy male participants. Each participant performed a series of arithmetic problems which presented in pseudorandom order. For each task, an arithmetic problem of true and false answers were presented in block stimulus paradigm within 30s followed by 30s rest. Participants made a judgement by pressing the handgrip button to indicate the wrong answer. Statistical Parametric mapping (SPM8) and Dynamic Causal Modelling (DCM10) were used to determine brain Functional Specialization (FS) and Effective Connectivity (EC). For each participant, a number of 56 EC models for PART I (Effects of Problem-Size) and 77 for PART II (Effects of arithmetic components) were constructed for each hemisphere to test the existence of couplings between Superior Parietal Lobule (SPL), insular, Inferior Occipital Gyrus (IOG) and Dorsolateral Prefrontal Cortex (DLPFC). These regions have been proved to be the key role in mental arithmetic and representing the fronto-parietal networks. The EC results show the existence of interactions between the regions of interest, but with different pattern of connectivity between hemispheres. In PART I, the SPL $\rightarrow$ DLPFC connection during single-digit MM task was the only one that was significant (Posterior probability, $P<0.90$ ) on the left hemisphere compared to other tasks. Meanwhile, the same connection, but on the right hemisphere was only significant during a MM task in PART II. This is due to both regions play important roles in fronto-parietal network, in which SPL is associated with number processing, while DLPFC is generally associated with working memory and task difficulty. Comparison study on MC and FR tasks where the answer can be directly retrieved from the memory, MM tasks were more complex, thus forcing the subject to retrieve, memorize and compute the answers at the same time. These findings indicated that both left and right hemispheres are involved in arithmetic processing. The results obtained from this study revealed that problem-size and arithmetic components do affect the EC between regions of interest despite their insignificantly difference in cortical activation.


#### Abstract

ABSTRAK

Pengetahuan arithmetik terdiri daripada beberapa komponen, termasuk kemahiran prosedural (pengiraan matematik mudah, (MC) dan pengetahuan matematik yang kompleks, (MM)), pemahaman konsep dan pengetahuan fakta (FR). Semua komponen ini telah menjadi rujukan asas apabila menjalankan kajian ke atas arithmetik kognitif menggunakan kaedah imbasan neuro. Kajian terdahulu telah membuktikan pengaktifan korteks fronto-parietal dengan keamatan yang berbeza semasa menjalankan pelbagai jenis masalah mental arithmetik. Beberapa kajian juga mencadangkan bahawa masalah saiz atau kesukaran boleh mempengaruhi pengaktifan fronto-parietal. Walau bagaimanapun, kurang penyelidikan yang telah dilakukan untuk menangani kesan masalah saiz dan pengiraan mathematik dari segi kehubungan efektif di dalam otak. Oleh itu, kajian ini dijalankan untuk menentukan kedua-dua kesan empirikal dari aspek pengkhususan kefungsian (FS) dan kehubungan efektik (EC) bagi rangkaian fronto-parietal semasa melakukan pengiraan matematik. Dalam kajian ini, imbasan pengimejan resonans magnet kefungsian (fMRI) dilakukan ke atas dua puluh dua ( $\mathrm{n}=22$ ) subjek lelaki yang sihat. Setiap subjek perlu melakukan satu siri arithmetik yang dipersembahkan dalam susunan rawak. Bagi setiap arithmetik, jawapan yang betul dan salah telah dipamerkan dalam paradigma blok rangsangan. Subjek membuat pengiraan secara mental dan menekan butang bagi menunjukkan jawapan yang salah. Pemetaan Statistik Berparamater (SPM8) dan Pemodelan Dinamik Penyebab (DCM10) digunakan untuk menentukan sifat FS dan EC otak. Untuk setiap subjek, sejumlah 56 model EC bagi BAHAGIAN I (Kesan Masalah Saiz) dan 77 model bagi BAHAGIAN II (Kesan Komponen Arithmetik) dibina untuk setiap hemisfera bagi menguji kewujudan gandingan di antara Lobus Superior Parietal (SPL), Girus Inferior Osipital (IOG), Insular dan Korteks Dorsolateral Prefrontal (DLPFC). Kawasan korteks ini terbukti memainkan peranan utama dalam mental arithmetik dan mewakili rangkaian korteks fronto-parietal. Keputusan bagi EC menunjukkan kewujudan interaksi antara kawasan otak yang terlibat, tetapi dengan corak EC yang berbeza di antara hemisfera. Di BAHAGIAN I, hanya SPL $\rightarrow$ DLPFC pada hemisfera kiri yang bererti (Kebarangkalian posterior,P<0.90), iaitu ketika satu-digit MM berbanding arithmetik lain. Sementara itu, EC yang sama tetapi pada hemisfera kanan menunjukan kehubungan bererti semasa MM dalam BAHAGIAN II. Ini kerana, kedua-dua korteks memainkan peranan penting dalam rangkaian fronto-parietal, di mana SPL dikaitkan dengan pemprosesan nombor, dan DLPFC amnya dikaitkan dengan ingatan kerja dan kesukaran tugas. Berbanding dengan MC dan FR, MM adalah lebih kompleks, di mana subjek perlu mendapatkan semula pengetahuan fakta sedia ada, menghafal dan mengira jawapan pada masa yang sama. Penemuan menunjukkan bahawa kedua-dua hemisfera otak terlibat dalam pemprosesan arithmetik. Hasil kajian menunjukkan bahawa masalah saiz dan komponen arithmetik mempengaruhi EC antara kawasan pengaktifan dalam korteks fronto-parietal meskipun pengaktifan mempamerkan perbezaan yang tidak ketara


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## LIST OF ABBREVIATIONS

| ACC | - | Anterior Cingulate Cortex |
| :--- | :--- | :--- |
| AG | - | Angular Gyrus |
| AIC | - | Akaike's Information Criterion |
| BIC | - | Bayes's Information Criterion |
| BMA | - | Bayesian Model Averaging |
| BMS | - | Bayesian Model Selection |
| BOLD | - | Blood-Oxygenated Level Dependent |
| DCM | - | Dynamic Causal Modelling |
| DICOM | - | Digital Imaging and Communications in Medicine |
| DLPFC | - | Dorsolateral Prefrontal Cortex |
| DTI | - | Diffusion Tensor Imaging |
| EEG | - | Electroencephalography |
| EMG | - | Electromyography |
| EPI | - | Event Relanar Imaging Potential |
| ERP | - | Negative-free Energy |
| F | - | Fixed Effect Analysis |
| FFX | - | Functional Magnetic Resonance Imaging |
| fMRI | - | Functional Near-Infrared Spectroscopy |
| fNIRS | - | Field Of View |
| FOV | - | Facts Retrieval |
| FR | - | Family-Wise Error Rate |
| FWER | Inferior Occipital Gyrus |  |
| GCM | - | Independent Component Analysis |
| ICA | - | IOG |
| IOG | - | - |


| IPC | - | Intra-Parietal Cortex |
| :---: | :---: | :---: |
| MEG | - | Magnetoencephalography |
| MC | - | Simple Mathematics Computation |
| MD | - | Complex Mathematics Computation |
| MPR | - | Multi-Planar Reconstruction |
| MRI | - | Magnetic Resonance Imaging |
| MS | - | Multiple Sclerosis |
| MVAR | - | Multivariate Autoregressive Model |
| NMR | - | Nuclear Magnetic Resonance |
| PCA | - | Principal Component Analysis |
| PET | - | Positron Emission Tomography |
| PMA | - | Premotor Association Area |
| PPM | - | Posterior Probability Map |
| PSC | - | Percent Signal Change |
| RFT | - | Random Field Theory |
| RFX | - | Random Effect Analysis |
| ROIs | - | Region Of Interests |
| SEM | - | Structural Equation Modelling |
| SMA | - | Supplementary Motor Area |
| SPL | - | Superior Parietal Lobule |
| SPM | - | Statistical Parametric Mapping |
| TCM | - | Triple Code Model |
| VBM | - | Voxel-Based Morphometry |
| VFG | - | Visual Fusiform Gyrus |
| VOI | - | Voxel Of Interest |
| WICA | - | Within-condition Inter-regional Covariance Analysis |

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## CHAPTER 1

## INTRODUCTION

In the first chapter, the scenario of mathematics in daily life as well as how neuroimaging research bridge the gap between neuroscience and mathematics education will be explained briefly under the first section of research background. Then, the problem statements of this work will be defined thoroughly followed by the contribution of this research in section three. In section four, a few research questions are listed. And as the objectives were lined up, the research scope will be drawn up as the boundary of the present study. At the end of this chapter, the research hypothesis is composed.

### 1.1 Introduction

One of the basic elements of mathematics are numbers. We use numbers almost daily to measure ordinary things such as height and weight, calculate daily expenses and even to estimate the education loan. In addition, numbers are used in various types of operations in mathematics such as ranking, counting and comparing quantities. Meanwhile, arithmetic knowledge comprises of several components, such as factual
knowledge, procedural skill, and conceptual understanding (Gilmore, 2006; Long, 2011). Generally, conceptual knowledge involves "knowing how", which is composed of a network of relationships among pieces of information. Problems that are involved an inversed transformation (e.g. $10+4-4=$ ?) are used to measure the conceptual knowledge of arithmetic between addition and subtraction.

On the other hand, procedural knowledge can be defined as "knowing-how-to", whereby composed of sets of procedures executed in a specific sequence and it primarily involves a standard arithmetic problem (e.g. 10+8-3=?). Meanwhile, the cognitive number processing of arithmetic fact retrieval was proposed by McCloskey since three decades ago by verbally testing on dyscalculia patients (Dagenbach and McCloskey,1992; McCloskey et al.,1985). McCloskey (1985) also proposed that calculation involving multiplication and division encompassed of several calculationspecific processes.

These processes postulate the mechanisms for (1) retrieval of arithmetic facts (e.g., $6 \times 7=42$ ), ( 2 ) execution of calculation procedures, and (3) comprehension of operation signs (e.g.,,$+ x$ ) or words (e.g.,plus). In conclusion, mental calculation can be described as a complex task that delineate an important component of higher order cognition that involves several fundamental cognitive activities. The specific processes involve covert production of numbers, execution of a calculated operation which is represented by arithmetic facts from a memorized table and storing data in working memory for further operations. This specific process is shown in Figure 1.1. Aschraft (1992) in his review journal entitled, "Cognitive arithmetic: A review of data and theory" has proposed four important empirical effects which can influence the organization of a person's knowledge of numbers and mathematics in the memory and the processes that can enable this knowledge to be accessed and applied in different settings. These engage the problem size/difficulty effects, error effects, relatedness effects and strategies of processing.

The problem-size effect has been described as a well-known phenomenon in mental arithmetic, in which the problem difficulty increases with the numerical size of the operands (Ashcraft, 1992). This shows that, the problem size effect is the most basic empirical effect in mental arithmetic studies, whereby as problems grow larger,
the process becomes more difficult. Taking this as a hypothesis, the first part of the current study was designed to observe the effect of problem size/difficulty on cortical activation as well as to how the information travel ipsilaterally between the regions. An inaugural study of mental arithmetic using neuroimaging modalities has been conducted about three decades ago by Roland and Friberg (1985). They measured the regional changes in cerebral blood flow, which was induced by different cognitive tasks, using single-photon emission tomography.


Figure 1.1 The cognitive process on calculation-specific processing for the mental arithmetic (Dagenbach and McCloskey, 1992; McCloskey et al., 1985)

A serial of subtraction problems used in arithmetic tasks has specifically activated the Angular Gyrus (AG) bilaterally. Upon this study, a number of significant neuropsychological studies have evolved over the last three decades, which concentrated on the cognitive mapping of number processing and calculation (Arsalidou and Taylor, 2011; Zamarian et al., 2009). This shows that in the last 30
years, the rapid development of brain imaging techniques has helped the identification of the neuro-anatomical brain areas supporting arithmetic functions more precisely.

Networks of brain regions are responsible for the different functions executed during arithmetic problem solving. However, based on the number of steps they require, this arithmetic decision has posed various numbers of cognitive demands, differently. Numerous neuroimaging studies focus on arithmetic processing using single-step arithmetic problems (e.g., $3+4,5-2$ ) which are composed of one-digit or a combination of one and two-digit numbers (Fehr et al., 2007). Besides, it also includes the manipulation of numbers in successive operations (e.g., $4-3+6$ ) (V Menon et al., 2000) or even solving integration problems (Ambady et al., 2009). Evidence from neuroimaging studies indicates that the neural substrate of arithmetic processes are composed of the cingulate cortex, parietal cortex (including intra-parietal sulcus and inferior parietal lobule), lateral prefrontal cortex, insula and occipital cortex (Eickhoff et al., 2009; Kong et al., 2005). These areas are proved to be the main region in mental arithmetic.

The involvement of prefrontal cortex in the brain activity has been linked to one of the cognitive functions such as working memory (Yoo et al., 2012), with considerable emphasis on its role in monitoring or manipulating information, as required in arithmetic tasks. Researchers who study numerical processing and computation perceive that difficult arithmetic tasks require more working memory resources than simple tasks (Fehr et al., 2007; Kong et al., 2005).

Practically, it is suggested that the three parietal circuits (Stanislas Dehaene et al., 2003a) are involved in arithmetic processing. The bilateral horizontal intraparietal sulcus/superior parietal lobule, which is one of the three parietal circuits is thought to be the domain-specific for arithmetical processing. Meanwhile, the other two regions, involving the left AG verbal system and the posterior parietal attention system are most probably shared with other cognitive domains. One of these parietal, which is the superior parietal lobule (SPL) bolster the attention, spatial working memory and visuospatial processes.

These processes are pertinent to numerical processing. Specifically, these three parietal regions are also known as Triple Code Model (TCM) (Stanislas Dehaene et al., 2003a; Schmithorst and Brown, 2004). TCM model predicts that numbers are handled in three numerical surface format: (1) a visual Arabic code portrayed by a series of digits (brought on the activation of bilateral activity in inferior ventral occipital-temporal areas), (2) an analogic quantity and magnitude code (generated activity in the left inferior parietal areas, which underlies quantity and magnitude judgement) and (3) verbal code represented by words (prompted the initiation of the left perisylvian areas). Inside this system, simple-digit calculations can be tackled either through a direct route utilizing operands (e.g., $3 \times 5$ ) transcoded into a verbal code (three times five), which would inspire the rote memory of this operation, or through an indirect semantic route in which the operands represent quantities in which semantical meaningful manipulations can be performed.

As for subtraction problem, the indirect route is regularly taken when rote memory for a problem is inaccessible. Therefore, Dehaene and Cohen (1996; 2003a) contended that addition and multiplication depend generally on rote verbal memory (direct route), while subtraction depends for the most part on quantitative manipulations (indirect route), and that these two processes are reflected in the brain as two main cortical systems for calculations.

The involvement of frontal, central-parietal (or temporal) and parietal components have been described in numerous fMRI and ERP studies regarding addition and subtraction (Burbaud et al., 1999, 2000; Chochon et al., 1999; Cowell et al., 2000; S. Dehaene, 1999; Rickard et al., 2000; S M Rivera et al., 2005; StanescuCosson et al., 2000; L. Zago et al., 2001; Rocha et al., 2005). These widespread areas are included in arithmetic calculations and have been articulated in the previous studies. All authors have focused on both the left frontal and parietal areas as common and critical components of the arithmetic brain.

Some of these authors have recommended that the frontal regions are greatly involved in the more difficult calculations (Anderson et al., 2011; Bongard and Nieder, 2010; V. Menon et al., 2000; Rypma and D'Esposito, 1999; Yu et al., 2012) while the temporal region is generally hypothesized as having a bilateral distribution in
arithmetic computations, its activity is depicted as for the most part reliant on the type of calculation and problem size (Chochon et al., 1999; Cowell et al., 2000; S. Dehaene, 1999; Rickard et al., 2000; Stanescu-Cosson et al., 2000). Whereas, some authors have additionally alluded to other visual and verbal components related to arithmetic calculation (Cowell et al., 2000; S. Dehaene, 1999).

In concomitant to the increasing numbers of cortical activation study on mental arithmetic, the focus began to shift from blobology to connectivity. The aboriginal study of connectivity underlying number processing has been conducted by Yiyuang Tang and colleagues (2006). They have developed a functional connectivity of visuopremotor association network of simple addition task by comparing the effects of culture on brain activities. The focus on connectivity underlying different types of mental arithmetic began to increase in a small number despite the increasing numbers of connectivity studies in other areas. This research study has been extensively used to observe and discover brain connectivity focusing more on cognitive brain function such as language, auditory, visual, working memory etc.

However, the effective connectivity underlying multiplication network has been discovered for the past five years using Multivariate Granger Causality method (Krueger et al., 2011). Other brain connectivity underlying mental arithmetic were also conducted using DTI fiber tractography (Tsang et al., 2009), functional connectivity (Emerson and Cantlon, 2012; Park et al., 2013), and Fiber Tract (Open DX)(Klein et al., 2013).

Despite the importance of number processing in daily life and numerous findings on the development of its underlying brain regions, little is known about the effective connectivity underlying number processing between fronto-parietal networks, unlike in other basic cognitive functions, such as auditory system. The ability to observe brain functions as well as brain connectivity that exists between the network during number processing via mental calculation enables us to study the changes in circuitry during different complexity (effect of problem size) and different operators (procedural knowledge vs. fact retrieval), thus it may help to improve effective teaching strategies.

In addition, neuroscience approach may add to open deliberation by demonstrating that diverse neural circuits are included in the procedural knowledge and the retrieval of arithmetic facts as an aftereffect of the present study upheld by past discoveries. Therefore, to the author's knowledge, this is the first study giving proof that different complexity (problem size) and different successive operators significantly affect the fronto-parietal networks attained from functional Magnetic Resonance Imaging (fMRI) and Dynamic Causal Modelling (DCM) Analysis.

In this study, the functional specialization, activation intensity and effective connectivity for two mental arithmetic tasks encompasses the effect of problem-size (Results presented in Chapter 4: Effect of Problem Size) and the impact of arithmetic components (Results presented in Chapter 5: Effect of Arithmetic Knowledge) were conducted with twenty-two healthy participants. The functional specialization of the brain is based on the tasks involving different types of arithmetic problems given during the fMRI experiments. The fMRI data collected from this study will be analysed using Statistical Parametric Mapping (SPM) method.

SPM is a statistical methodology of Random Field Theory (RFT) that is used to make inferences about the topological features of statistical processes that are continuous functions of space or time and it is typically used to distinguish regionally particular impact in neuroimaging data to describe functional anatomy and diseaserelated changes (K. J. Friston et al., 1995). Functional data obtained from the experiment needs to be matched with standard anatomical space to associate the activation with the brain regions. Next, parameter estimation will be conducted using general linear model to specify the statistical models which represent the haemodynamic response of the tasks. Then, the statistical inference is made based on the Gaussian Random Field Theory and this step is known as functional specialization.

To obtain the analysis of functional integration, the percentage signal change (PSC) analysis on the region of interests (ROIs) is conducted. The purpose of this analysis is to explore the data, which involves various types of stimulus or tasks (Mohammad, 2012). Besides, the analyses will focus only on the brain regions that are dependent on the given task. The signal changes are defined as relative signal changes
in the brain regions. To analyse the effective connectivity between regions of interest, the dynamic causal modelling (DCM) is used.

DCM posits a causal model whereby neuronal activity in a given region induces changes in neuronal activity in different areas, by means of interregional connections, and in its own particular activity, through self-connections. A DCM is fitted to data by tuning the neuro-dynamic and haemodynamic parameters so as to minimize the error between predicted and observed fMRI time series. A few models are constructed by specializing the activated brain regions, which refer to the different mental arithmetic tasks. These models will be fitted with the measured BOLD signals. Therefore, the intrinsic connection between regions and how they are influenced by the experiment-context can be obtained.

Then, all models are analysed using Bayesian Model Selection (BMS) and Bayesian Model Averaging (BMA). BMS is a powerful technique in deciding the most likely, among a set of contending hypotheses of the mechanisms that generate the observed data (K.E. Stephan et al., 2009). Whereas, BMA is used to produce the average network structure of the winning models (Will D. Penny et al., 2010).

### 1.2 Problem Statements

Either in natural phenomenon or technologies arising around us, mathematics is used extensively as it expresses itself everywhere and in almost every facet of life. This scenario supports the fact that mathematics is the language of science and engineering. Besides, mathematics is defined as the science of numbers, and as a matter of fact human being did not invent mathematics concepts instead, we discovered them. Uniquely, the language of mathematics is in numbers. Not Malay, English or Japanese language.

In the present study, both specialization and integration analysis proposed in neuroimaging study were conducted in order to deeply understand how the brain regions are involved in mathematics working memory response between frontoparietal networks under different approach of mathematical tasks. Study on functional specialization focuses on the region activated during mental arithmetic task has been reported numerously. Detail on neuroscience of learning arithmetic based on fMRI evidence has been summarized by Zamarian et al. (2009). However, little is known about functional integration analysis which focuses more on "How the brain interact to one another during certain tasks (Stevens, 2009)" especially in the field of learning arithmetic.

The distribution of previous studies, which analysed and observed the mental arithmetic performance on brain activity through different fields of study is shown in Figure 1.2. Therefore, the present study will clearly discuss activated fronto-parietal network across different mathematical tasks and how the brain works during these tasks. Thus, this present study will bridge the gap between experimental psychological study of working memory with education (mental arithmetic) and neuroscience study (the fronto-parietal activation and connectivity) in order to draw out a better understanding of mental arithmetic processing within the brain on normal subjects.

At the end of this study, the outcome can be made as a reference baseline of disconnected brain network of dyscalculia (the disability of doing mathematics) subjects while doing the same mental arithmetic study. Thus, a clear distinction of fronto-parietal connectivity between normal and dyscalculia subjects could be justified in the future. Therefore, it is hoped that through this study, researchers could develop more effective teaching method of basic mathematics learning (e.g., mental arithmetic) for dyscalculia subject as well as lower performance subjects in such a way that can help them survive in the real world since basic mathematical calculation is a vital component in life.


Figure 1.2 The distribution of previous study conducted on mental arithmetic tasks through different fields of study

### 1.3 Research Questions

There are several questions that arose during this study:
a. Which part of the brain is involved in arithmetic processing?
b. How are these regions connected with one another during arithmetic processing?
c. Does problem-size/ difficulty affect arithmetic processing? And how?

### 1.4 Research Objectives

The present research study was conducted in order to bridge the gap that exists between neuroscience and education. Generally, this study was conducted in order to analyse the functional specialization and integration under two different chapters. Chapter 4 focuses on the effects of problem-size, which is indicated by single- and double-digit problems, while Chapter 5 concentrates on the effects of mathematics components during mental arithmetic tasks, which encompass procedural and factual knowledge. There are several objectives that this study hopes to achieve, which are:
i. Identifying the functional specialization in the brain regions
ii. Comparing the percentage of signal change on the four regions of interests (ROIs), which comprise of IOG, SPL, DLPFC and insular for both left and right hemispheres
iii. Determining the structure of effective connectivity between the regions by contrariety the connection between SPL to IOG, DLPFC and Insular
iv. Constructing the effective connectivity model for each case, which can explain how the ROIs interact with one another during computation of mental arithmetic tasks

### 1.5 Research Scope

The scope or boundaries of this research study comprise of a few aspects which include the modality used as well as the image format. Besides, it also explains the boundaries of subjects selected in this research study. For the modality, the present
study only uses functional Magnetic Resonance Imaging (fMRI) as one of the neuroimaging modalities. Other modalities will not be used in this study. Therefore, the image used in this study will be a set of fMRI time-series, which results from the brain scanning during mental calculation tasks given during the experimental session.

Subjects or participants selected for this research study were Diploma students from the University Teknologi Malaysia (Kuala Lumpur Campus) with 11 years of education background. Although the nature and origin of gender differences in mathematics remains unclear, variability in performance tends to be higher in males (Feingold, 1992; Geary, 1998). Consequently, to reduce variability in the present study, only male students are involved in this study. The whole analyses conducted in this study were based on Statistical Parametric Mapping (SPM8) and Dynamic Causal Modelling (DCM). This research study also focuses only on mental calculation involving four basic arithmetic operations, which is one of the basic cognitive functions.

### 1.6 Significance of Study

Compared to the developed countries in the world, research on brain activities using non-invasive, radiation - free technology of fMRI is still new and growing in the developing countries, especially in Malaysia. This functional Magnetic Resonance Imaging (fMRI) is utilized to assess how the brain's blood flow and oxygen change, in the specific areas, react to certain stimuli. It can outline how language and speech are acquired, how pain and other emotions are handled, how numbers are prepared and accumulated and a bunch of other brain functions.

From this study, we will be able to give new and important information on various types of mental arithmetic problems and how the regions of interest interact
with one another during mathematics computation tasks, which is unable to be obtained simply through behavioural analysis or clinical tasks. In addition, the understanding on how mental arithmetic computes under different cases can be done in this study in terms of functional specialization and integration aspects.

Robust studies have been conducted on simple addition and multiplication tasks involving single-digits and the brain regions that specialize in number processing have been identified. However, not much research has been done to tackle the doubledigit problems and effects of math computation of more than one operator. Therefore, this study was conducted in order to give new insight of math computation via fMRI data analysis. Moreover, from this study, it is hoped that the construction of effective connectivity model for each case will give some basic understanding of the connectivity patterns between SPL and the fronto-parietal system, including insular, DLPFC and IOG.

Besides, analysis of mental arithmetic using fMRI can give additional info to the current psychological analysis in terms of brain neurophysiology mechanism as well as functional specialization and integration. By comparing with previous findings, this study contributes in the use of dynamic causal modelling (DCM), together with Bayesian Model Selection (BMS) and Bayesian Model Averaging (BMA) to scrutinize brain connectivity between regions during mental arithmetic tasks. In addition, this is the first study that investigates the intrinsic connectivity constructed between regions that is modulated by problem-size effects as well as the arithmetic operators (the mathematic computation and fact retrieval).

### 1.7 Thesis Organization

This thesis is organized into six chapters.

Chapter 1 provides the introduction of the project, which contains a brief executive summary information where the scope of the project is also discussed. It also presents the motivation and objectives of the research done. Several facts about the previous work by other researchers are also touched.

Chapter 2 contains literature review and details about the information and scope of the research. It also illustrates prior knowledge concerning the brain and its cognitive functions as well as modern neuroimaging technologies used to examine and discover the neural network underlying these functions.

Chapter 3 discusses briefly on the methodology for the project in every research stages. These include the five main phases which is data collection, data pre-processing, model specification, inference on the brain activation and the intrinsic connectivity analysis.

Chapter 4 presents the results obtained from Statistical Parametric Mapping (SPM8) and Dynamic Causal Modelling (DCM10) analysis during the problem size tasks. These include Fixed Effect Analysis (FFX), Random Effect Analysis (RFX), Region of Interest (ROIs) and Effective Connectivity (EC) analysis. Discussion will be made based on i) FFX and RFX analysis as well as ROI analysis involving functional specialization of activated brain regions and ii) effective connectivity analysis demonstrating the interaction between regions of interest (ROIs) during single- and double-digit.

Chapter 5 presents the results obtained from SPM8 and DCM10 analysis carried out to discover the effect of Arithmetic Component. Discussion will be made based on both functional specialization of activated brain regions as well as the effective connectivity analysis demonstrating the interaction between ROIs during math computation and fact retrieval.

Chapter 6 concludes the thesis with a review of the objectives and their fulfilment, a summary of the work that has been accomplished, and recommended future work.

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