



UNIVERSITI PUTRA MALAYSIA
UNIVERSITY ENTRY SELECTION FRAMEWORK USING
RULE-BASED AND BACK-PROPAGATION

SITTI SYARAH BINTI MAHARANI

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**UNIVERSITY ENTRY SELECTION FRAMEWORK USING RULE-BASED
AND BACK-PROPAGATION**

By

SITTI SYARAH BINTI MAHARANI

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

October 2015

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DEDICATIONS

In the name of Allah, Most Gracious, Most Merciful This thesis is dedicated to:

*Husband & Son
Father & Mother
Family & Friends*



Abstract of thesis presented to the Senate of Universiti Putra Malaysia in
fulfilment of the requirement for the Degree of Master of Science

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October 2015

Chair : Razali Yaakob, PhD
Faculty : Computer Science and Information Technology

Processing thousands of applications can be a challenging task, especially when the applicant does not consider the university requirements and their qualification. The selection officer will have to check the program requirements and calculate the merit score of the applicants. This process is based on rules determined by the Ministry of Education and the institution will have to select the qualified applicants among thousands of applications.

In recent years, several student selection methods have been proposed using the fuzzy multiple decision making and decision trees. These approaches have produced high accuracy and good detection rates on closed domain university data. However, current selection procedure requires the admission officers to manually evaluate the applications and match the applicants' qualifications with the program they applied. Because the selection process is tedious and very prone to mistakes, a comprehensive approach to detect and identify qualified applicants for university enrollment is highly desired.

In this work, a student selection framework using rule-based and back-propagation neural network is presented. Two processes are involved in this work; the first phase known as pre-processing uses rule-based for checking the university requirements, merit calculation and data conversion to serve as input for the next phase. The second phase uses back-propagation neural network model to evaluate the qualified candidates for admission to particular programs. This means only selected data of the qualified applicants from the first phase will be sent to the next phase for further processing. The dataset consists of 3,790 datasets from Universiti Pendidikan Sultan Idris.

The experiments have shown that the proposed method of ruled-based and back-propagation neural network produced better performance, where the framework has successfully been implemented and validated with the average

performance of more than 95% accuracy for student selection across all sets of the test data.



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

**KERANGKA KERJA PEMILIHAN KEMASUKAN UNIVERSITI
MENGUNAKAN KAEDAH BERASASKAN PERATURAN DAN
PERAMBATAN-BALIK**

Oleh

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Memproses beribu-ribu permohonan boleh menjadi tugas yang mencabar, terutamanya apabila pemohon tidak mengambil kira syarat am universiti dan kelayakan mereka. Pegawai pendaftaran akan menyemak syarat khas program dan mengira skor merit pemohon. Proses ini adalah berdasarkan kepada peraturan yang ditetapkan oleh Kementerian Pelajaran Malaysia dan institusi yang akan memilih pemohon yang layak di kalangan beribu-ribu permohonan.

Pada tahun-tahun kebelakangan ini, beberapa kaedah pemilihan pelajar telah dicadangkan menggunakan kaedah pembuat keputusan kabur pelbagai dan pepokok keputusan. Pendekatan ini menghasilkan ketepatan yang tinggi dan kadar pengesanan yang baik pada data dari domain tertutup universiti tertentu. Walau bagaimanapun, prosedur pemilihan yang digunakan kini memerlukan pegawai kemasukan untuk menilai secara manual permohonan dan cuba untuk menyesuaikan kelayakan pemohon dengan program yang mereka mohon. Oleh kerana ia adalah satu proses yang rumit dan sangat terdedah kepada kesilapan, pendekatan yang menyeluruh untuk mengesan dan mengenal pasti pemohon yang layak untuk pendaftaran universiti amat diinginkan.

Dalam kajian ini, kerangka kerja pemilihan pelajar menggunakan kaedah berasaskan peraturan dan rangkaian neural perambatan-balik dibentangkan. Dua proses terlibat dalam kerja ini; fasa pertama menggunakan kaedah berasaskan -peraturan untuk memeriksa syarat am universiti, pengiraan merit dan penukaran nilai data sebagai input bagi fasa seterusnya. Fasa kedua menggunakan kaedah rangkaian neural perambatan-balik untuk menilai calon-calon yang layak untuk kemasukan ke program-program tertentu. Ini bermakna hanya data pemohon yang layak daripada fasa pertama sahaja akan dihantar ke fasa seterusnya. Set data mengandungi 3,790 dataset dari Universiti Pendidikan Sultan Idris.

Hasil kajian telah menunjukkan bahawa kaedah yang dicadangkan iaitu kaedah berasaskan peraturan dan rangkaian perambatan-balik menghasilkan prestasi yang lebih baik, di mana rangka kerja ini telah berjaya dilaksanakan dan disahkan dengan purata ketepatan melebihi 95% bagi pemilihan pelajar bagi semua set pengujian data.



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Lastly, I would like to thank to all those who had lent me a helping hand in allowing me to materialize this study.

I certify that a Thesis Examination Committee has met on 22 October 2015 to conduct the final examination of Sitti Syarah binti Maharani on her thesis entitled "University Entry Selection Framework Using Rule-Based and Back-Propagation " in accordance with the Universities and University Colleges Act 1971 and the Constitution of the Universiti Putra Malaysia [P.U.(A) 106] 15 March 1998. The Committee recommends that the student be awarded the Master of Science.

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LIST OF ABBREVIATIONS

ABE	Agent Building Environment
AI	Artificial Intelligent
ANN	Artificial Neural Network
CASCEP	Computer-Aided in Choosing Education Program
EST	English Language for Science and Technology
IPTA	Institusi Pengajian Tinggi Awam
IPTS	Institute Pengajian Tinggi Swasta
KPT	Kementerian Pengajian Tinggi
MADM	Multiple Attribute Decision Making
MCE	Malaysian Certificate of Education
MLP	MultiLayer Perceptron
MOE	Ministry of Education
MOHE	Ministry of Higher Education
PMR	Penilaian Menengah Rendah
RAISE	Reusable Agent Software Environment
RSAU	Recommender System of Admission to University
SMJK	Sekolah Menengah Jenis Kebangsaan
SMK	Sekolah Menengah Kebangsaan
SMKA	Sekolah Menengah Kebangsaan Agama
SPM	Sijil Pelajaran Malaysia
SRP	Sijil Rendah Pelajaran
STPM	Sijil Tinggi Pelajaran Malaysia
UITM	Universiti Teknologi Mara
UKM	Universiti Kebangsaan Malaysia
UNET	University Entry Selection Framework Using Rule-Based and Back-Propagation
UNITAR	Universiti Tun Abd Razak
UNITEN	Universiti Tenaga Nasional
UPM	Universiti Putra Malaysia
UPSI	Universiti Pendidikan Sultan Idris
UPSR	Ujian Penilaian Sekolah Rendah
UPU	Unit Pusat Universiti
UTM	Universiti Teknologi Malaysia

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MCE	Malaysian Certificate of Education
MLP	MultiLayer Perceptron
MOE	Ministry of Education
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PMR	Penilaian Menengah Rendah
RAISE	Reusable Agent Software Environment
RSAU	Recommender System of Admission to University
SMJK	Sekolah Menengah Jenis Kebangsaan
SMK	Sekolah Menengah Kebangsaan
SMKA	Sekolah Menengah Kebangsaan Agama
SPM	Sijil Pelajaran Malaysia
SRP	Sijil Rendah Pelajaran
STPM	Sijil Tinggi Pelajaran Malaysia
UITM	Universiti Teknologi Mara
UKM	Universiti Kebangsaan Malaysia
UNET	University Entry Selection Framework Using Rule-Based and Back-Propagation
UNITAR	Universiti Tun Abd Razak
UNITEN	Universiti Tenaga Nasional
UPM	Universiti Putra Malaysia
UPSI	Universiti Pendidikan Sultan Idris
UPSR	Ujian Penilaian Sekolah Rendah
UPU	Unit Pusat Universiti
UTM	Universiti Teknologi Malaysia

CHAPTER 1

INTRODUCTION

This chapter forms the introduction to the thesis. Discussions begin with the overview of tertiary education in Malaysia with highlights in generic problems of student selection in university admission.

1.1 Background

Education has always been highly valued and is often regarded as a secure passport for promising future. Education is also viewed as a critical factor in contributing to the long-term economic well-being of a country. The government of any country has long realized the importance of maximizing the potential of each individual future graduate in an educational system.

In Malaysia, the young adults are eligible to pursue tertiary education after finishing their secondary education at the age of 17 (Yassin, 2015). At this juncture, they have two choices to pursue their studies, whether to enrol in a public higher education institutions (IPTA) such as Universiti Putra Malaysia (UPM) and Universiti Perguruan Sultan Idris (UPSI) or in a private higher education institutions (IPTs) such as Universiti Tun Abdul Razak (UNITAR) and Universiti Tenaga Nasional (UNITEN).

Since 2011, the intakes for all IPTA across Malaysia have been standardized to commence in September of every year (Nordin, 2011). The universities, nonetheless, all have their own standards in selecting applicants to enroll in their respective institution. Because of the wide range of programs and programs offered, coupled with the enormous number of applicants, it is difficult for officers in the admission department to conduct a transparent and consistent selection.

Applications to universities in Malaysia are coordinated by the Ministry of Higher Education (MOHE), whereby all applications will go through Unit Pusat Universiti (UPU) (MOHE, 2011). The final decision on the university selection is made by UPU based on the applicants' qualification and their choice of place of study. UPU is known to place the applicant in one of the university in the applicants' list based on their academic eligibility.

In order to enrol in the public university, the applicants should have at least a credit in Bahasa Melayu and at least credits in four other subjects (Mustafah, 2011). For instance, applicants who wish to pursue the Diploma in Science is required to meet the minimum requirements of the university as well as the specific subject requirements for the program. The additional requirements are set to ensure applicants will be able to perform well throughout the program. The requirements are shown in Figure 1.1 and Figure 1.2 (MOE, 2014).

UNIVERSITY REQUIREMENTS
<p style="text-align: center;">Minimum Requirements</p> <ol style="list-style-type: none"> 1. Malaysian citizen, and 2. Pass in Malaysia education certificate or equivalent with at least five (5) credits including Malay Language.

Figure 1.1: University Requirement

(i) Program Education (ii) Code Program (iii) Duration of Study	Minimum Requirements
<p>DIPLOMA OF SCIENCE A2000</p> <p>06 Semester</p> <p>Open to Category A: Science Stream</p> <p>Diploma Program at UPSI</p>	<p>Fulfill the University Requirements and Program Requirement</p> <p>Obtain at least a credit (Grade C) at SPM level in any two (2) of the following subjects:</p> <ul style="list-style-type: none"> • Chemistry • Biology • Physics • Mathematics • Additional Mathematics • Engineering of Drawings • School of Electrical and Electronic Engineering • School of Civil Engineering • Mechanical of Engineering <p style="text-align: center;">and</p> <p>At least pass or Grade E at SPM level in the following subjects:</p> <p>English Language</p>

1.2: Program Requirement for Diploma in Science

Although the list of applicants being considered for admission is ranked based on their merit (Mustafah, 2011), the process of selecting the most qualified candidates for a particular program is carried out manually due to additional criteria to be considered in programs selected. In this research, Universiti Perguruan Sultan Idris (UPSI) has been chosen as the case study to validate a framework on student selection and focus on diploma level only. UPSI also offers tertiary education at the diploma level, hence the applicants have just completed the *Sijil Pelajaran Malaysia* (SPM). SPM holders carry a variety of subjects at the SPM level. Appendix A detail out the complete requirements of the programs offered by UPSI.

1.1 Problem Statement

Selecting qualified applicants to enter one educational institution is not an easy task because it involves a lot of processes and requirements. The selection committee has to consider several things including the applicants' qualifications, the programs requirements and the quota for each program. In fact, each institute and faculty also have their own quota.

Since 2003, the intake of students to universities will no longer be based on the 55 per cent quota places allocated to Bumiputera while the other 45 percent to non-Bumiputera (Mustafah, 2011). This method does not have a category of grades for excellent candidates and difficult to select suitable applicants among them because the aggregate A in SPM result increasing every year.

The similar issue encountered by selection officer each university, if allocating applicants for one program in a faculty from a particular university is difficult enough; imagine if it involves hundreds of programs from several faculties and universities. It will definitely involve many people and will take a very long time, not to mention the labour cost. Sometimes the selection officer overlook the qualification of applicants. In addition, there are high chances for the applicants to overlook or even ignore the minimum requirements of each program, therefore making the selection process more cumbersome.

Based on this situation, the government introduced the meritocracy system that opens equal opportunities to students of various races based on student ability. The system of meritocracy use 90% academic marks and 10% cocurriculum marks. The determination of the merits carried out regardless of religion, race, descent, land, town and so on (Mustafah, 2011).

Several techniques have been proposed to facilitate the selection process in order to identify the most suitable programs for each applicant. The techniques include the fuzzy MADM (Mohamad et al., 2005) and rule-based reasoning (Mohd et al., 2009) developed to assist applicant for selecting a university program based on academic results. The fuzzy method (Mokhtar et al., 2011) also used to recommend suitable programs based on high school results. The neural network model (Adewale et al., 2007) and (Wabwoba and Mwakondo, 2011) used for student selection based on applicants' qualification and programs requirements. Lei et al. (2015) examines parameters based on academic

performance, personal Information and socio-Economic. In Fong et al. (2009) paper, neural network with decision tree classifier used to assist selection officer to analyze data from secondary school result.

Nonetheless, there are still room for improvement to enhance the accuracy in selection of the right program for each applicant. This is because to date, none of the related works have considered the meritocracy marks at the secondary school level, while the extra curricula involvement is known to be a strong indication for a qualified applicant (MOE, 2014). The calculation of merit is based on two components; 90% Academic Marks and 10% Co-curriculum Marks from school assessment (Nordin, 2011). The academic merit and cocurriculum merit will be combined for the final merit score with a total mark of 100%. In this study, the process only focuses in the calculation of academic marks while the co-curriculum marks (Appendix C) has been evaluated by their secondary schools.

1.2 Objectives of the Study

The main objective of this research is to **propose a framework for student selection in university admission to assist the selection officer in verifying and selecting student application**. The detailed framework is as follow:

1. To include the ruled-based method for pre-processing phase to check the university requirements, merit calculation and data conversion.
2. To use the back-propagation algorithm to evaluate the qualified candidates for admission to a particular programs.

1.3 Scope of the Study

This research focuses only on admission process by analyzing candidates' applications in programs offered by Universiti Pendidikan Sultan Idris (UPSI). It is developed using a rule-based and back-propagation neural network method and is able to list suitable programs for diploma level based on SPM results and co-curriculum marks.

1.4 Organization of Thesis

This thesis is organized in accordance to the standard structure of thesis dissertations for Universiti Putra Malaysia. The thesis is inherently divided into six chapters as follows:

Chapter 1 – Introduction. This chapter introduces the degree of importance on student selection and its impact to the public. Awareness on current selection issues forms the problem statement and the research objective.

Chapter 2 – Literature Review. This chapter reviews related studies on the fundamental knowledge related to the subject such as the selection systems, criteria of selection, university requirements, machine learning techniques such as hybrid, neural network and other related techniques.

Chapter 3 – Research Methodology. This chapter presents an overview of research steps, which comprise of problem identification, dataset preparation, design of the proposed method, implementation of proposed method, and finally experiment and analysis.

Chapter 4 – University Entry Selection (UNET). This chapter introduces the back-propagation neural network to evaluate student applications for university admission. The output of the proposed approach is a list of all programs qualified by the applicants with ranking merits for each student application.

Chapter 5 – Experiments and Results. This chapter presents the data source, metric of performance evaluation, as well as the experimental process flow that formed the experiments. In addition, this chapter also presents a analysis between the proposed technique other approaches.

Chapter 6 – Conclusion and Future Works. This chapter concludes the research with some recommendations for future development.

BIBLIOGRAPHY

- Adewale, O. S., Adebiji, A. B. and Solanke, O. O. 2007. Web-based Neural Network Model for University Undergraduate Admission Selection and Placement. *Pacific Journal of Science and Technology* 8 (2): 367–384.
- Admir, M., Ivan, B., Ivan, D. and Nina, B. 2014. Decision Tree Based Students' Grades Analysis . In *IEEE 12th International Symposium on Applied Machine Intelligence and Informatics*, 133 – 136. IEEE.
- Ansari, Shafi, Ahmad and Shah. 2011. Diagnosis of Liver Disease Induced By Hepatitis Virus using Artificial Neural Networks. In *IEEE 6th International Conference on Emerging Technology*, 8 – 12. IEEE.
- Arora, D., Neville, S. and Li, K. F. 2013. Mining WiFi Data for Business Intelligence . In *P2P, Parallel, Grid, Cloud and Internet Computing (3 PGCIC), 2013 Eighth International Conference*, 394–398. IEEE.
- Aura, P., Kaisa, P. and Pirjo, P. 2011. Women and Higher Engineering Education-Choosing Ones Degree Program. In *41th ASEE/IEEE Frontiers in Education Conference*, T2H–1 – T2H–6. IEEE.
- Baohua, L. 2007. Design and realization of college students comprehensive quality evaluation system based on data mining . In *Guilin: Guangxi Normal University*. in Chinese.
- Baumann, T. 2014. Decision Tree Usage for Incremental Parametric Speech Synthesis . In *IEEE International Conference on Acoustic, Speech and Signal Processing (ICASSP)*, 3819 – 3823. IEEE.
- Breiman, L. 1999. Random forests . In *Technical Report, Statistics Department, University of California, Berkeley*.
- Breiman, L. 2001. Random forests . In *Machine Learning*, 5–32. Springer Link.
- Busch, C. and Grob, M. H. 2008. Interactive Neural Network Texture Analysis and Visualization for Surface Reconstruction in Medical Imaging. *Computer Graphic Forum* 12 (3): 49–60.
- Caiyun, D. and Shouning, Q. 2004. Data mining and its application in university education system. In *Journal of Jinan University (Sci. and Tech.)*, 65–68. Chinese.
- Chaiyan, J., Chaichan, P. and Boonlert, S. 2012. Selection of Proper Activation Functions in Back-propagation neural networks algorithm for Transformer Internal Fault Locations . In *Soft Computing and Intelligent Systems (SCIS) and 13th International Symposium on Advanced Intelligent Systems (ISIS), 2012 Joint 6th International Conference*, 1487 – 1492. IEEE.

- Cleopatra, Veloutsou, J. L. and Paton, R. A. 2004. University selection: information requirements and important . In *The International Journal of Educational Management*, 160–171.
- Danyang, C. 2006. Data mining applying and studying in educational administration system . In *Beijing: North China University of Technology*. in Chinese.
- Das, S., Dahiya, S. and Bharadwaj, A. 2014. An online software for decision tree classification and visualization using c4.5 algorithm (ODTC) . In *Computing for Sustainable Global Development (INDIACom), 2014 International Conference*, 962 – 965. New Delhi: IEEE.
- Dawi, A. H. 2006. *Pentecorian of Sociology and Education*. Perak, Malaysia: Quantum Books.
- Dazhuo, Z., Minqiang, L. and Hongcan, Y. 2008. An Efficient Similarity Search For Financial Multivariate Time Series . In *Wireless Communications, Networking and Mobile Computing, 2008. WiCOM '08. 4th International Conference*, 1 – 4. Dalian: IEEE.
- Dominic, P. B., Chrisina, D. and Sin, W. L. 2009. Snap-Drift Neural Network for Selecting Student Feedback . In *Neural Networks, 2009. IJCNN 2009. International Joint Conference*, 391 – 398. Atlanta, GA: IEEE.
- Du, K. 2011. Clustering: A neural network approach . In *Neural Networks*, 89–107.
- Elshazly, H., Azar, A., El-Korany, A. and Hassanien, A. 2013. Hybrid system for lymphatic diseases diagnosis . In *Advances in Computing, Communications and Informatics (ICACCI), 2013 International Conference*, 343–347. Mysore: IEEE.
- Farhood, zdemir, F., Yksel, K. A., Akgl, C. B. and Ercil, A. 2013. A Decision Forest Based Feature Selection Framework for Action Recognition from RGB-Depth Cameras . In *Signal Processing and Communications Applications Conference (SIU)*, 1 – 4. IEEE.
- Farrah, Diana, S. B., Noraizan, A., Norizan, A., Sobariah, Awang, M., Zuraidah, Abd, M., Norzanah, Md, N. and Megawati, O. 2013. Factors in Selection of a Malaysian University . In *IEEE Business Engineering and Industrial Applications Colloquium (BEIAC)*, 399 – 403. IEEE.
- Feng, Q. X. 2006. The research on application of data mining in management of students grades . In *Liaoning Technical University*. in Chinese.
- Fillali., F., Boubetra., A., Maza., S. and Sakhraoui, R. 2013. Enhancement of Decision Strategies Selection Under Uncertainty Using Discrete Simulation . In *Modelling Symposium (EMS), 2013 European*, 189 – 194. Manchester: IEEE.

- Fong, S., Si, Y. W. and Biuk-aghai, R. P. 2009. Applying a Hybrid Model of Neural Network and Decision Tree Classifier for Predicting University Admission. In *Information, Communications and Signal Processing, 2009. ICICS 2009. 7th International Conference*, 1 – 5. IEEE.
- Geethalakshmi, S. N., Subashini, P. and Ramya, S. 2011. A Study on Detection and Classification of Underwater Mines Using Neural Networks. *International Journal of Soft Computing and Engineering* 1 (5).
- Goltz, E., Arcoverde, G., de Aguiar, D., Rudorff, B. and Maeda, E. 2009. Data mining by decision tree for object oriented classification of the sugar cane cut kinds . In *Geoscience and Remote Sensing Symposium, 2009 IEEE International, IGARSS 2009*, 405–408. IEEE.
- Grana, C., Borghesani, D. and R.Cucchiara. 2010. Optimized block-based connected components labeling with decision trees . In *IEEE Transactions on Image Processing*, 1596 – 1609. IEEE.
- Guang-Bin, H., Qun-Yu, Z., Moa, K., Chee-Kheong, S., P, S. and N, S. 2006. Can Threshold Networks be Trained Directly? . In *IEEE Transactions On Circuits And SystemsII: EXPRESS BRIEFS*, 187–191. IEEE.
- Han, Y., Xiao, Y. and Zhang, Z. 2007. Application of association rules mining to arrangement of courses in universities . In *Microcomputer Applications*, 54–57. in Chinese.
- Hanife, G., Halil, I. B. and Erdal, I. 2013. The Estimation of Students Academic Success by Data Mining Methods . In *Machine Learning and Applications (ICMLA), 2013 12th International Conference*, 535–539. Miami, FL: IEEE.
- Holt, R. 1993. Very simple classification rules perform well on most commonly used datasets . In *Machine Learning*, 69–90.
- Huang, C. Y., Shi., S.-F., Liu., Z. and Huang, W. P. 2013. Application of Data Mining on HPLC Fingerprints of Szechwan Lovage Rhizome Analysis . In *Mobile Ad-hoc and Sensor Networks (MSN), 2013 IEEE Ninth International Conference*, 560–564. IEEE.
- Huang, L.-J. 2006. Recognizing Real Customer in E-Supply Chain Based on Softm Neural Network and Corresponding Marketing Strategies. In *Machine Learning and Cybernetics, 2006 International Conference*, 1592 – 1597. IEEE.
- Huo, Z. 2014. Data mining: The investment value analysis on China's bank stocks based on residual income model . In *Electronics, Computer and Applications, 2014 IEEE Workshop*, 326–329. IEEE.
- Ibrahim, S. 2002, IPTA Admission – Selected Program Must be Accurate, <http://www.utusan.com.my/>, retrieved Aug 8th 2002.

- Intrator, N. and Gold, J. I. 1993. Three-Dimensional Object Recognition Using an Unsupervised BCM Network: The Usefulness of Distinguishing Features. *Neural Computation* 5: 61–74.
- Jiang, X.-F. 2012. The Research on Sales Forecasting Based on Rapid BP Neural Network. In *Computer Science and Information Processing (CSIP), 2012 International Conference*, 1239 – 1241. IEEE.
- Jiawei, H. and Micheline, K. 2006. Data mining: concepts and techniques. In *(Second Edition)*, Elsevier Inc. Amazon.
- Jixin, S. 2013. Data Mining in High-Frequency Financial Data - Long Memory Test in Chinese Agriculture Futures' Market . In *Computational and Information Sciences (ICCIS), 2013 Fifth International Conference*, 340–343. Shiyang: IEEE.
- John, G. H. 1995. Estimating Continuous Distributions in Bayesian Classifiers . In *In Proceedings of the Eleventh Conference on Uncertainty in Artificial Intelligence, Morgan Kaufmann Publishers, San Mateo*.
- Joze, H. and Drew, M. 2014. Exemplar-Based Color Constancy and Multiple Illumination . In *Pattern Analysis and Machine Intelligence, IEEE Transactions*, 860 – 873. IEEE.
- Kasinathan, M., Saidhar, R. B., M. N. and Swaminathan, P. 2009. An Artificial Neural network Approach for the Discordance Sensor Data Validation for SCRAM Parameters. In *Advancements in Nuclear Instrumentation Measurement Methods and their Applications (ANIMMA), 2009 First International Conference*, 1 – 5. IEEE.
- Kayim, Sari and Akgul, C. 2013. Facial feature selection for gender recognition based on random decision forests . In *Signal Processing and Communications Applications Conference (SIU)*, 1–4. Haspolat: IEEE.
- Kohavi, R. 1995. The Power of Decision Tables . In *8th European Conference on Machine Learning*, 174–189.
- Kumar, S. A. and Vijayalakshmi. 2012. Mining Of Student Academic Evaluation Records in Higher Education. In *Recent Advances in Computing and Software Systems (RACSS), 2012 International Conference*, 67 – 70. Chennai: IEEE.
- Kyrillidis., A. and Zouzias, A. 2014. Non-Uniform Feature Sampling for Decision Tree Ensembles . In *IEEE International Conference on Acoustic, Speech and Signal Processing (ICASSP)*, 4548 – 4552. IEEE.
- Lau, Sear, H. 2009. Higher Education Marketing Concern: Factors Influencing Malaysian Students Intention to Study at Higher Learning Institution . In *Master Thesis, University of Malaya, Kuala Lumpur*.

- Lawrence, S., Giles, C. L., Tsoi, A. C. and Back, A. D. 2000. Face Recognition: A Convolutional Neural Network Approach. In *IEEE Transaction on Neural Networks*, 98 – 113. IEEE.
- Lei, C., and Kin Fun, L. 2015. Academic Performance Predictors. In *Advanced Information Networking and Applications Workshops (WAINA), 2015 IEEE 29th International Conference*, 577 – 581. IEEE.
- Li, Z. and Ziyao, L. 2012. Commercial Bank Customer Satisfaction Research Based on Kano Model and BP Neural Network. In *IEEE Symposium on Robotics and Applications (ISRA)*, 657 – 659. IEEE.
- Liu, Y., S. H. and Li, P. 2007. *Word Naming and Psycholinguistic Norms in Chinese.* , vol. 39. Springer-Verlag.
- LLC, T. 2015, WEKA The University of WAIKATO.
- Lu, J., Plataniotis, K., Venetsanopoulos, A. and Li, S. 2006. Ensemble-based discriminant learning with boosting for face recognition . In *IEEE Transaction on Neural Networks*, 166–178. IEEE.
- Mansi, G. and Shivani, G. 2015. A Model for Predicting the Eligibility for Placement of Students Using Data Mining Technique. In *International Conference on Computing, Communication and Automation (ICCCA2015)*, 114–117. IEEE.
- Martin, T. H., Howard, B. D. and Orlando, D. J. 2002. An introduction to the use of neural networks in control systems . In *International Journal Of Robust And Nonlinear Control*, 959–985.
- Mattei, N., Dodson, T., Guerin, J. T., Goldsmith, J. and Mazur, J. M. 2014. Lessons Learned from Development of a Software Tool to Support Academic Advising . In *Proceedings of 2014 Zone 1 Conference of the American Society for Engineering Education (ASEE Zone 1)*.
- Mendes, J., Sousa, N. and Araujo, R. 2011. Adaptive Predictive Control with Recurrent Fuzzy Neural Network for Industrial Processes. In *Emerging Technologies and Factory Automation (ETFA), 2011 IEEE 16th Conference*, 1 – 8. IEEE.
- Menglin, L. 2007. Clustering based on particle swarm optimization algorithm and its application in the students grade management . In *Jinan: Shandong Normal University*. in Chinese.
- Michael, N. 2005. Artificial Intelligence . In *A Guide to Intelligent Systems, 2nd edn. Addison Wesley*.
- MOE. 2014, Unit Pusat Universiti (UPU), <https://online.moe.gov.my>, retrieved Aug 8th 2014.

- Mohamad, D., Alumni, S. A. and Rashid, J. A. 2005. A University Program Selection Expert System for SPM-Leavers using Fuzzy Multiple Attribute Decision Making Method. In *National Conference on Programming Science*, 156–165.
- Mohammad, N. M., Linkon, C. and Md, S. K. 2012. Students Dropout Prediction for Intelligent System from Tertiary Level in Developing Country . In *Informatics, Electronics and Vision (ICIEV), 2012 International Conference*, 113 – 118. Dhaka: IEEE.
- Mohar, Y. and et al. 2008. A study of factors influencing the selection of higher education institution. In *Unitar e-Journal*. Unitar.
- Mohd, F., Noor, N. M. M. and Yusof, Y. 2009. Computer-Aided in Choosing Education Program (CASCEP). *Internation Journal of Soft Computing* 4 (2): 85–94.
- Mohd, N. 2013, Panduan Kemasukan ke IPTA 2012/2013.
- MOHE. 2011, Unit Pusat Universiti (UPU), <https://online.mohe.gov.my>, retrieved Aug 8th 2011.
- Mokhtar, I. A., Zulkifli, N. A. and Shaffie, S. S. 2011. Intelligent Higher Institution Student Selection System. In *Computer Applications and Industrial Electronics (ICCAIE), 2011 IEEE International Conference*, 396–401. IEEE.
- Moret, B. M. E. 1982. Decision Trees and Diagrams . In *ACM Computing Surveys (CSUR)*.
- Munkongsujarit, S., Schweinfart, W., Iskin, I., Colon, R., Tanatammatorn, N., Phopoonsak, N. and Almobarak, A. 2009. Decision Model for a Place to Live at PSU: The Case of International Graduate Students . In *Management of Engineering and Technology, 2009. PICMET 2009. Portland International Conference*, 513 – 534. Portland, OR: IEEE.
- Mustafah, R. 2011, Pekeliling SPM, <http://upu.mohe.gov.my/portal/arkib/1213PEKELILINGSPM.pdf>, retrieved October 28 th.
- Nahato, K. B., Harichandran, K. N. and Arputharaj, K. 2015. Knowledge Mining from Clinical Datasets Using Rough Sets and Backpropagation Neural Network . In *Computational and Mathematical Methods in Medicine*.
- Najarian, K. 2001. On learning of sigmoid neural networks . In *Complexity*, vol. 6, no. 4.
- Negin, F., Ozdemir, F., Yuksel, K., Akgul, C. and Ercil, A. 2013. A decision forest based feature selection framework for action recognition from RGBdepth cameras . In *Signal Processing and Communications Applications Conference (SIU)*, 1–4. IEEE.

- Nigrin, A. 1996. *Data Mining with Neural Networks: Solving Business Problem: From Application Development to Decision Support*. McGraw-Hill, NY.
- Nordin, M. K. 2010. *Admission Handbook to Public Higher Education Institution: SPM Leavers/Equivalent Program, Academic Session 2010/2011 , July Intake*. Department of Higher Education.
- Nordin, M. K. 2011, New academic year in September, <http://www.thestar.com.my/education/story.asp?sec=education&file=/2011/1/16/education/7793531>.
- Nordin, M. K. 2013. *Admission Handbook to Education in Malaysia: Carta Sistem Pendidikan Malaysia*. Department of Higher Education Malaysia.
- Nordin, T., Ahmad, R. and Rahim, R. A. 2007. *Membina Pelajar Cemerlang Evolusi Pembelajaran Sepanjang Hayat*. Johor, Malaysia: Universiti Teknologi Malaysia.
- Oladokun, V. O., Adebajo, A. T. and Charles-Owaba, O. E. 2008. Predicting Student Academic Performance using Artificial Neural Network: A Case Study of an Engineering Course. *The Pacific Journal of Science and Technology* 9 (1): 72–79.
- Ozkan, N. 2010. Digging Social Networks by Mashups to Support Recruitment and Selection Functions in University Student Intake Process. In *Business Applications of Social Network Analysis (BASNA), 2010 IEEE International Workshop*, 1 – 8. IEEE.
- Panigrahi, S. and Behera, H. S. 2013. Effect of Normalization Techniques on Univariate Time Series Forecasting using Evolutionary Higher Order Neural Network . In *International Journal of Engineering and Advanced Technology (IJEAT)*.
- Perez, R. L. and Gil, P. G. 2010. Unconstrained Handwritten Word Recognition Using a Combination of Neural Networks. In *Proceedings of the World Congress on Engineering and Computer Science 2010 (WCECS)*, 1 – 4.
- Pinilla, J. M. C., S, W. and Aguiar, P. 2006. Engineering Students DecisionMaking Process: The Role of Aspirations . In *Technology Management for the Global Future, 2006. PICMET 2006*, 489 – 497. Istanbul: IEEE.
- Popescu, B. 2004. Ensemble learning for prediction . In *Stanford, CA, USA Stanford University*.
- Qing, Zhou., Y., Zheng. and Chao, M. 2015. Predicting Students' Performance of an Offline Course from Their Online Behaviors. In *International Conference on Computing, Communication and Automation (ICCCA2015)*, 70 – 73. IEEE.
- Quinlan, R. 1993. Programs for Machine Learning . In *Morgan Kaufmann Publishers, San Mateo, CA*.

- Rafael T. S., , Marques., O., C., G. T. F. and M, R. 2014. Evaluation of Classifiers to a Childhood Pneumonia Computer-aided Diagnosis System . In *Computer-Based Medical Systems (CBMS), 2014 IEEE 27th International Symposium*, 477 – 478. IEEE.
- Ramon, V., Paz, L. and Francisco, M. 2014. Wind speed estimation using multilayer perceptron . In *Energy Conversion and Management*, 1 – 9.
- Ravindra, Pal, S. 2012. Design of an optimal multi-layer neural network for eigenfaces based face recognition . In *Recent Research in Science and Technology*, 24–32. IEEE.
- Reddy, U., Arock, M. and Reddy, A. 2013. A particle swarm optimization solution for challenging planted(l, d)-Motif problem . In *Computational Intelligence in Bioinformatics and Computational Biology (CIBCB), 2013 IEEE Symposium*, 222 – 229. IEEE.
- Reinhard, H., Eirikur, A. and Helmut, B. 2014. Neighborhood Selection for Thresholding-Based Subspace Clustering . In *IEEE International Conference on Acoustic, Speech and Signal Processing (ICASSP)*.
- Rodriguez, P., Drenkow, N., DeMenthon, D., Koterba, Z., Kauffman, K., Cornish, D., Paulhamus, B. and Vogelstein, R. 2014. Selection of universal features for image classification . In *Applications of Computer Vision (WACV), 2014 IEEE Winter Conference*, 355–362. Steamboat Springs, CO: IEEE.
- Rokia, J. and Sigit, P. J. 2012. Intelligent Decision Support System for Degree Selection Using AHP Technique. In *Computer and Communication Engineering (ICCCE), 2012 International Conference*, 642 – 647. Kuala Lumpur: IEEE.
- Rozanc, I. and Slivnik, B. 2014. Analysis of elective courses selection in post-Bologna programmes . In *Information and Communication Technology, Electronics and Microelectronics (MIPRO), 2014 37th International Convention*, 744 – 749. Opatija: IEEE.
- Rumelhart, D., Hinton, G. and Williams, R. 1986. Learning internal representations by back-propagating errors . In *In D. Rumelhart and J. McClelland, editors, Parallel Distributed Processing: Explorations in the Microstructure of Cognition*, MIT Press, Cambridge, 318–362.
- Rumelhart, D. E. 1993. Strategies for developing effective neural network applications . In *Neural Networks, 1993. IJCNN '93-Nagoya. Proceedings of 1993 International Joint Conference*.
- Schumacher, H. and Sevcik, K. C. 1976. The synthetic approach to decision table conversion . In *Commun. ACM*.

- Shamim., A., Hussain., H. and Shaikh, M. U. 2010. A Framework for Generation of Rules from Decision Tree and Decision Table . In *Information and Emerging Technologies (ICIET), 2010 International Conference*, 1 – 6. IEEE.
- Shangguang, W., Cuncun, F., Ching, H. ., Sun., Q. and Yang, F. 2014. A Vertical Handoff Method via Self-Selection Decision Tree for Internet of Vehicles . In *Systems Journal, IEEE*, 1 – 10. IEEE.
- Simon, H. 2008. *Neural Networks and Learning Machines* . In *Hoboken, NJ, USA: Wiley*.
- Skapura, D. M. 1995. *Building Neural Networks*. AddisonWesly, New York.
- Songphon, M., Willi, S., Ibrahim, I. and Rafael, C. 2009. Decision Model for a Place to Live at PSU: The Case of International Graduate Students. In *Management of Engineering and Technology, 2009. PICMET 2009. Portland International Conference*, 513 – 534. Portland: IEEE.
- Sousa, A., Prudencio, R., Soares, C. and Ludermir, T. 2013. Active selection of training instances for a random forest meta-learner . In *Neural Networks (IJCNN), The 2013 International Joint Conference*, 1–7. Dallas, TX: IEEE.
- Sylvain, R. 2002. Nearest Neighbor With Generalization . In *Christchurch, New Zealand..*
- Taha., R. A., Choi., B. C., Chuengparsitporn., P., Cutar., A., Gu., Q. and Phan, K. 2007. Application of Hierarchical Decision Modeling for Selection of Laptop . In *Management of Engineering and Technology, Portland International Center*, 1160 – 1175. Portland, OR: IEEE.
- Taruna, Mrinal and Pandey. 2014. An Empirical Analysis of Classification Techniques for Predicting Academic Performance . In *Advance Computing Conference (IACC), 2014 IEEE International*, 523 – 528. Gurgaon: IEEE.
- Tingting, W. 2007. Research on the application of data warehouse and data mining in the analysis of students academic achievements . In *Wuhan University of Science and Technology*. in Chinese.
- Tung, k. W., Shian, C. H. and Ying, R. M. 2007. Improving ANN Classification Accuracy for the Identification of Students with LDs through Evolutionary Computation . In *Evolutionary Computation, 2007. CEC 2007. IEEE Congress*, 4358 – 4364. Singapore: IEEE.
- Vanmali, M., Last, M. and Kandel, A. 2002. Using a Neural Network in the Software Testing Process. *International Journal of Intelligent System* 17: 8–12.
- Vapnik, V. 1995. The Nature of Statistical Learning Theory . In *SVM. Springer-Verleg*.

- Victor, G., Wadhwa., N., Sandeep, G. and Sanjeev, G. 2012. Applying Fuzzy MADM Approach for the Selection of Technical Institution. In *Industrial Engineering and Engineering Management (IEEM), 2012 IEEE International Conference*, 1405 – 1408. IEEE.
- Wabwoba, F. and Mwakondo, F. M. 2011. Students Selection for University Course Admission at the Joint Admissions Board (Kenya) Using Trained Neural Networks. *Journal of Information Technology Education* 10: 333 – 347.
- Wan, J., Tan, L., Tan, T. and Lai, S. 2007. The data mining based design and realization of students online course-selecting system. In *Computer Development and Applications*, 19–21. in Chinese.
- Wang, C. 2007. Data mining in teaching appraisal applied research. In *Jinan: Shandong Normal University*. in Chinese.
- Wang, G., Wang, H., ., J. L. and Fan, W. 2014. Mining Knowledge Sharing Processes in Online Discussion Forums . In *System Sciences (HICSS), 2014 47th Hawaii International Conference*, 3898–3907. Waikoloa, HI: IEEE.
- Wang, J., Huiyuan, S., Shasha, D. and Piel, W. 2003. TreeRank: a similarity measure for nearest neighbor searching in phylogenetic databases . In *Scientific and Statistical Database Management, 2003. 15th International Conference*, 171 – 180. IEEE.
- Wang, Y. and Li, A. 2007. Applying association rules mining to credit management system. In *Sun Yatsen University Forum*, 165–167. in Chinese.
- Wei, C. 2006. Research and realization of digital campus based on data mining . In *Hangzhou: Zhejiang University*. in Chinese.
- Wets, G., Witlox, F., Timmeman, H. and Vanthienen, J. 1996. A Fuzzy Decision Table Approach for Business Site Selection . In *Fuzzy Systems, 1996., Proceedings of the Fifth IEEE International Conference*, 1605 – 1610. IEEE.
- Wu, F. 2010. Apply Data Mining to Students Choosing Teachers Under Complete Credit Hour. In *Education Technology and Computer Science (ETCS), 2010 Second International Workshop*, 606 – 609. Wuhan: IEEE.
- Wu, T., Lin, C. and Weng, R. 2004. Probability estimates for multi-class classification by pairwise coupling . In *Journal of Machine Learning Research*, 975–1005.
- Xianxiang, C. 2007. Apply data mining to evaluating college students comprehensive quality. In *Guiyang: Guizhou University*. in Chinese.
- Xin, G. H. and Shao, H. X. 2003. Process neural network with time varied input and output functions and its applications . In *Journal of software*, 764=769.
- Xindong, W., Vipin, K. and Ross, J. 2008. Top 10 algorithms in data mining . In *Knowledge Information System*, 1–37. Springer Link.

- Xingui, H. and Shaohua, X. 2010. Process neural networks . In *Process Neural Networks: Theory and Applications, Advanced Topics in Science and Technology in China*.
- Xini, W. 2006. College student raise and career guidance research with data mining . In *Chengdu: Sothwest Jiaotong University*. in Chinese.
- Yaakob, R. 2008. Integration of A Best Population Pool and Social Learning: An Investigation in Game Playing . In *Phd Thesis, University of Nottingham*.
- Yadav, R. and Danvir Mandal, D. 2011. Optimization of Artificial Neural Network for Speaker Recognition using Particle Swarm Optimization. *International Journal of Soft Computing and Engineering* 1, Issue 3: 80–84.
- Yang, H., Bai, X., Zhou, J., Ren, P., Zhang, Z. and Cheng, J. 2014. Adaptive Object Retrieval with Kernel Reconstructive Hashing . In *Computer Vision and Pattern Recognition (CVPR), 2014 IEEE Conference*, 1955 – 1962. IEEE.
- Yassin, M. 2015, Pelajaran Menengah, <http://www.moe.gov.my/v/pelajaranmenengah>, retrieved July 17th 2015.
- Yihew, W. M. and Hua, A. Z. 2014. Nearest Neighbor Relay Selection with Adaptive Modulation for Improved Throughput and Scalability of Cooperative Wireless Networks . In *Intelligent and Advanced Systems (ICIAS), 2014 5th International Conference*, 1 – 5. IEEE.
- Yu, Y. 2011. Risk management Game Method of the Weapons Project based on BP Neural Network. In *Information Technology, Computer Engineering and Management Sciences (ICM), 2011 International Conference*, 113 – 117. IEEE.
- Yuta, N. and Kenichi, T. 2012. Knowledge Acquisition on Selection of Humanities and Sciences Using the Decision Tree . In *IEEE International Conference on Systems, Man, and Cybernetics*, 126 – 131. IEEE.
- Zhang, Tong. Chen, C. Z. J. 2014. Impact of ratio k on two-layer neural networks with dynamic optimal learning rate . In *Neural Networks (IJCNN), 2014 International Joint Conference on*, 2738–2742. IEEE.