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The clustering of high schools based on national and school examinations : A case study at Daerah Istimewa Yogyakarta Province (Conference Paper)

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#### Abstract

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The purpose of Indonesian National Examination for High School Students is to measure and assess students' knowledge and competence in particular subjects. The result is also going to be used as one of consideration for mapping Indonesia's national education quality. Aside from National Examination (NE), each school also conducts School Examination (SE). Both examinations are supposed to represent guality of education since the examinations measure the competence of the same students. However, the results of both examinations are not always linear [1]. In fact, the need of NE in Indonesian education is still being pro and cons among society. In order to identify whether NE and SE could be used to represent the quality of educations in Daerah Istimewa Yogyakarta Province, this paper describes the analysis of NE and SE score by performing data mining technique using Fuzzy C-Means clustering algorithm towards NE score and SE score independently. Furthermore, the clusters were analyzed using univariate Anova, Spearman correlation, and crosstabulation. The data used in this research are NE and SE scores of Natural Science Department and Social Science Department of all high schools in Daerah Istimewa Yogyakarta Province from academic year 2011/2012 to 2014/2015. The results of cluster analysis are three different clusters of NE in Natural Science Department, three different clusters of NE in Social Science Department, three different clusters of SE in Natural Science Department, and three different clusters of SE in Social Science Department for each year. The clusters are significantly separated. There is an opposite direction relationship between clusters of NE and SE. The relationship is weak which means there is no guarantee that a school which belongs to cluster-i of NE will be in the cluster-i of SE. Both for NE and SE memberships, only few members migrated from one cluster to another across years. The number of schools having the same cluster of NE and SE in each department varies from year to year, but generally less than 22%. The migrations of NE and SE cluster members from higher cluster to lower one and vice versa also vary. In addition, there is inconsistency clustering based on NE and SE. Since SE is not standardized and indeed is a formative test, there might be subjective aspects involved in grading the students. Therefore, if the government intends to map Indonesia's national education quality, national examination is more suitable than school examination for this purpose. © 2015 IEEE.

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Topic: Fuzzy Clustering | Fuzzy C-Mean Algorithm | Tsallis Entropy

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Anova Clustering (crosstab) (fuzzy c-means) (Indonesian National Examination

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I	Zulkardi Model peningkatan mutu pendidika komering ilir	an sma di kota prabumulih, kabupaten ogan ilir, dan kabupaten ogan	

(2011) Prosiding Seminar Nasional Hasil Penelitian Universitas Sriwijaya

5

Khairunnisa, R., Fajriya, H. Pengelompokan sekolah dasar berdasarkan rata-rata nilai ujian akhir sekolah berstandar nasional di setiap sd/mi di kota yogyakarta dengan menggunakan metode hierrarkhi clustering Naskah Seminar Kerja Praktek, FMIPA, UII Yogyakarta, 2013

6	Pakpahan, Juni, M. Pengelompokan sekolah menengah pertama berdasarkan rata-rata nilai ujian akhir nasional di kota binjai dengan analisis hierrarkhi clustering (2013) <i>Naskah Skripsi Universitas Negeri Medan</i>
7	Prihatiningtyas Analisis hasil ujian nasional tingkat sma di kabupaten banyumas menggunakan analisis cluster dan biplot (2011) <i>Skripsi Program Studi Statistika, Universitas Diponegoro, Semarang</i>
8	Habibi, Ahmad, A. Pemetaan un propinsi di indonesia melalui analisis klaster penggabungan pautan tunggal (2009) <i>Thesis JBPTITBPP, ITB Bandung, Indonesia</i> . Cited 2 times.
9	Fadhli Analisis kluster untuk pemetaan mutu pendidikan di aceh (2011) <i>Thesis Jurusan Matematika Universitas Gadjah Mada Yogyakarta</i> . Cited 2 times.
□ 10	(2009) Sjafrudin and Asep, Analisis Hasil Ujian Nasional Madrasah Tsanawiyah Tahun 2008
□ 11	Ruspini, E.H. A new approach to clustering (Open Access) (1969) <i>Information and Control</i> , 15 (1), pp. 22-32. Cited 1071 times. doi: 10.1016/S0019-9958(69)90591-9 View at Publisher
12	Dunn, J.C. A fuzzy relative of the ISODATA process and its use in detecting compact well- separated clusters (1973) <i>Journal of Cybernetics</i> , 3 (3), pp. 32-57. Cited 3740 times. doi: 10.1080/01969727308546046 View at Publisher
13	Bezdek, J.C., Ehrlich, R., Full, W. FCM: The fuzzy c-means clustering algorithm (1984) <i>Computers and Geosciences</i> , 10 (2-3), pp. 191-203. Cited 3161 times. doi: 10.1016/0098-3004(84)90020-7 View at Publisher
14	Lu, Y., Ma, T., Yin, C., Xie, X., Tian, W., Zhong, S. Implementation of the fuzzy c-means clustering algorithm in meteorological data (2013) <i>Int. J. Database Theory Appl</i> , 6 (6), pp. 1-18. Cited 41 times.

15	Luthfi, E.T. Fuzzy c-means untuk clustering data ( studi kasus: Performance mengajar dosen) (2007) <i>Semin. Nas. Teknol. 2007 (SNT) 2007 2007, No. November</i> , pp. 1-7. Cited 3 times.
□ 16	Susanto, S., Ernawati Pembagian kelas peserta kuliah berdasarkan fuzzy clustering dan partition coefficient and exponential separation ( pcaes ) index (2007) <i>J. Teknol. Ind</i> , 9 (2), pp. 143-154. APRIL
☐ 17	Karti, H.S. Pengelompokan Kabupaten /Kota di Provinsi SMA /SMK /MA dengan Metode C-Means dan Fuzzy C-Means (2013) <i>J. Sains Dan Seni Pomits</i> , 2 (2), pp. D-288.
18	Tan, PN., Steinbach, M., Kumar, V. (2005) <i>Introduction to Data Mining</i> . Cited 6109 times. First. Boston, MA, USA: Addison-Wesley Longman Publishing

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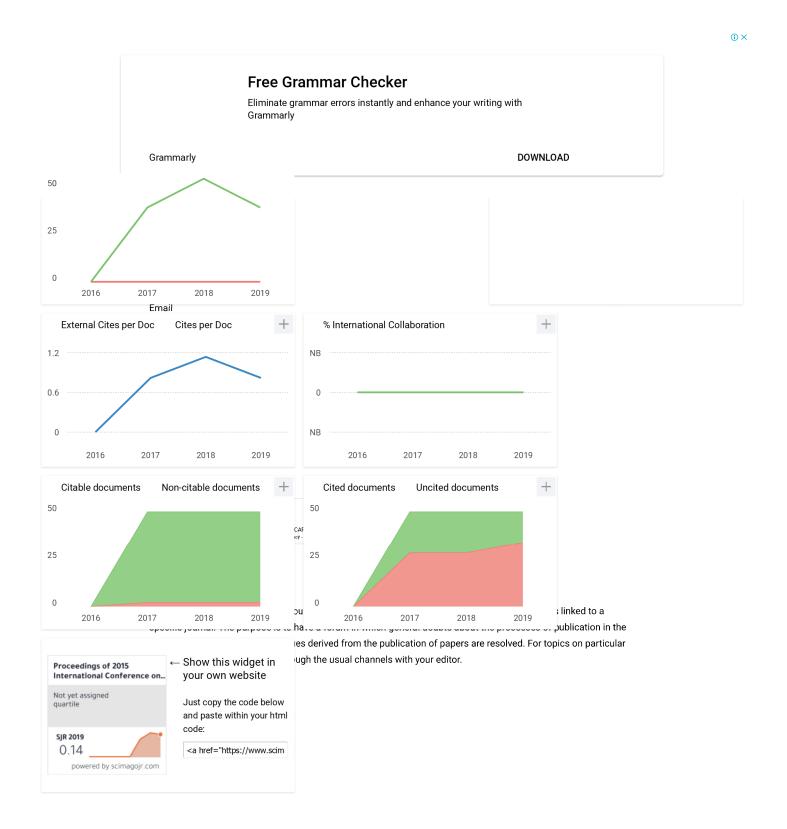
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#### Welcome to 2015 ICODSE.

It is a great pleasure for us to host the 2015 International Conference on Data and Software Engineering (ICODSE 2015) at Yogyakarta, Indonesia. The ICODSE2015 conference aims at uniting researchers and professionals in the domains of data and software engineering, presenting and discussing high-quality research results and outcomes in their fields. This year, the conference is organized by Department of Computer Science and Electronics, Faculty of Mathematics and Natural Sciences, Universitas Gadjah Mada. 2015 ICODSE is coorganized by Institut Teknologi Bandung and is also technically co-sponsored by IEEE Indonesia Section.

In total, we received 91 submissions of authors from 7 countries around the world. All submissions were peer-reviewed (blind) by at least three reviewers drawn from external reviewers and the committees, and as the result, 45 papers were accepted to be presented in this conference. These papers are in the proceedings of 2015 ICODSE.

Finally, as the General Chair of the Conference, I would like to express my deep appreciation to all members of the Steering Committee, Technical Programme Committee, Organizing Committee and Reviewers who have devoted their time and energy for the success of the event. For all participants, i hope you find the conference stimulating, fulfilling and enjoyable. I thank you for your support of ICoDSE2015 and your attendance, and wish you a pleasant experience in this conference and this beautiful city of Yogyakarta.

Aina Musdholifah General Chair of the ICoDSE2015

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# Contents

A Declarative Query Language Based on Speech Act Theory for Web Systems Bambang Purnomosidi D. P., Lukito Edi Nugroho, Paulus Insap Santosa, Widyawan	1
A Domain - Specific Languange for Automatic Generation of Checkers Ryan Ignatius Hadiwijaya, M.M. Inggriani Liem	7
<b>A Metamodel for Disaster Risk Models</b> Wawan Hendriawan Nur, Fazat Nur Azizah, Saiful Akbar	13
A Proposal for a Quality Model for E-Government Website Bayu Hendradjaya, Rina Praptini	19
A Proposal of Software Maintainability Model using Code Smell Measurement Billy C. Wagey, Bayu Hendradjaya, M. Sukrisno Mardiyanto	25
A Study of Disaster Situation Management Using Mobile Technology in Yogyakarta Yohanes Sigit Purnomo WP, Theresia Devi Indriasari, Kusworo Anindito, irvan	31
Agile Software Engineering in UCD Sowmya Dhandapani	37
Aircraft Anomaly Detection Using Algoritmic Model and Data Model Trained on FQDA Data Alvin Megatroika, Maulahikmah Galinium, Adhiguna Mahendra, Neno Ruseno	42
Analysis of K-means Algorithm For VM Allocation in Cloud Computing Bramantyo Adrian, Lukman Heryawan	48
Automated Data Consistency Checking Using SBVR Case Study : Academic Data in a University Vania Natali , Inggriani Liem	54
<b>Confidentiality and Privacy Information Security Risk Assessment for Android-Based Mobile Devices</b> <i>Irwan, Yudistira Asnar, Bayu Hendradjaya</i>	60
<b>Database Analysis and Design Learning Tool Based on Problem/Project- Based Learning</b> <i>Ruth Nattassha, Fazat Nur Azizah</i>	66
Decision Tree Modeling for Predicting Research Productivity of University Faculty Members	70

Arfika Nurhudatiana, Adilla Anggraeni

<b>Defending One-Time Pad Cryptosystems from Denial-of-Service Attacks</b> <i>Marc W. Abel, Soon M. Chung</i>	77
Deriving Labeled Training Data for Topic Link Detection by Alternating Words Marc W. Abel, Soon M. Chung	83
<b>Developing A Game for Preschoolers: What Character, Emotion and Reward will Tend to Hack Preschoolers?</b> <i>Endah Sudarmilah, Adhi Susanto, Ridi Ferdiana,Neila Ramdhani</i>	89
<b>Distributed Replicated Block Device (DRDB)</b> Mardhani Riasetiawan, Ahmad Ashari, Irwan Endrayanto	93
<b>Document Clustering using Sequential Pattern(SP)</b> Dini Rahmawati, G.A. Putri Saptawati, Yani Widyan	98
Fake Smile Detection Using Linear Support Vector Machine I Gede Aris Gunadi, Agus Harjoko, Retantyo Wardoyo, Neila Ramdhani	103
Feedback Fraud Detection in Online Marketplace System based on Fusion Approach Muhammad Harits Shalahuddin Adil Haqqi Elfahmi, Gusti Ayu Putri Saptawati	108
<b>Grid-based Histogram of Oriented Optical Flow for Analyzing Movements on</b> <b>Video Data</b> <i>Achmad Solichin, Agus Harjoko, Agfianto Eko Putra</i>	114
Implementation and Validation of Business Process Deviation Detection Framework Budi J. Achmadi, Bayu Hendradjaya, Wikan D. Sunindyo	120
<b>Implementation of an Optical Character Reader (OCR) for Bengali Language</b> <i>Muhammed Tawfiq Chowdhury, Md. Saiful Islam, Baijed Hossain Bipul, Md. Khalilur</i> <i>Rhaman</i>	126
Integration of HTML Tables in Web Pages Memen Akbar, Fazat Nur Azizah, G. A. Putri Saptawat	132
Interoperability Model for egoverment Service Based On Adaptive Ontology I Wayan Ordiyasa, Lukito Edi Nugroho, Paulus Insap Santosa, Ridi Ferdiana, Wahyudi Kumorotomo	137
Knowledge Discovery On Drilling Data To Predict Potential Gold Deposit Gusti Ayu Putri Saptawati , Gusti Ngurah Mega Nata	143
<b>Multiple mapreduce and Derivative Projected Database : New Approach for</b> <b>Supporting prefixspan Scalability</b> <i>Puspita Nurul Sabrina, G.A. Putri Saptawati</i>	148

<b>Optimation Weather Parameters Influencing Rainfall Prediction Using</b> <b>Adaptive Network-Based Fuzzy Inference System(ANFIS) And Linear</b> <b>Regression</b> <i>Devi Munandar</i>	154
<b>Optimization Of Real-Time Multiple-Face Detection In The Classroom Using</b> <b>Adaboost Algorithm</b> <i>Hadi Santoso, Agus Harjoko, Agfianto Eko Putra</i>	160
<b>Personal Health Care Framework for Children</b> Nina Sevani	166
Personality Classification Based on Twitter Text Using Naïve Bayes, KNN and SVM Bayu Yudha Pratama, Riyanarto Sarno	170
Preliminary Diagnosis of Pulmonary Tuberculosis Using Ensemble Method Rusdah, Edi Winarko, Retantyo Wardoyo	175
<b>Prototype of Moving Object Visualization Engine</b> Yani Widyani, Elia Dolaciho Bangun, Hira Laksmiwati, Rickard Elsen	181
<b>Public Facilities Recommendation System based on Structured and Unstructured Data Extraction from Multi-Channel Data Sources</b> <i>Alifa Nurani Putri, Saiful Akbar, Wikan Danar Sunindyo</i>	185
Quantifying visual attention and visually induced motion sickness during day- night driving and sleep deprivation Sunu Wibirama, Titis Wijayanto, Hanung A. Nugroho, Muhammad Bahit, Mumtaz N. Winadi	191
Risk-Level Assessment System on Bengawan Solo River Basin Flood Prone Areas Using Analytic Hierarchy Process and Natural Breaks Study Case: East Java Haris Rahadianto, Arna Fariza, Jauari Akhmad Nur Hasim	195
River Flood Spreading Prediction System Using Cellular Automata (Case Study Bengawan Solo River) Riza Budi Prasetya, Arna Fariza, Jauari Akhmad Nur Hasim, Achmad Basuki	201
Service Orchestration using Enterprise Service Bus for Real-Time Government Executive Dashboard System Kabul Kurniawan, Ahmad Ashari	207
Software Architecture for Social Media Data Analytics Anggi Perwitasari, Saiful Akbar, G.A. Putri Saptawati	213

<b>Source Code Generator for Automating Business Rule Implementation</b> Nisa'ul Hafidhoh, Inggriani Liem, Fazat Nur Azizah	219
Spatio-Temporal Queries for Disaster Information in Spatem dimas Indonesia Hira Laksmiwati, Yani Widyani	225
<b>The Clustering of High Schools Based on National and School Examinations</b> <b>A Case Study at Daerah Istimewa Yogyakarta Province</b> <i>Paulina H. Prima Rosa, Ridowati Gunawan, Ignatius Aris Dwiatmoko</i>	231
<b>The Role of Ontology in Big Data</b> <b>Integrating social data and organizational data for efficient decision-making</b> <i>Tengku Adil Tengku Izhar, Mohammad Fazli Baharuddin, Torab Torabi, Bernady</i> <i>O. Apduhan</i>	237
<b>Traffic Lights Detection and Recognition based on Color Segmentation and Circle Hough Transform</b> <i>Dwi H. Widyantoro, Kevin I. Saputra</i>	243
Unit Test Code Generator for Lua Programming Language Junno Tantra Pratama Wibowo, Bayu Hendradjaya, Yani Widyan	247

### **Author Index**

Abel, Marc	14, 15
Achmadi, Budi	22
Adrian, Bramantyo	9
Akbar, Memen	24
Akbar, Saiful	3, 34, 39
Anggraeni, Adilla	13
Anindito, Kusworo	6
Ashari, Ahmad	38
Asnar, Yudistira	11
Azizah, Fazat	3, 12, 24, 40
Bahit, Muhammad	35
Bangun, Elia	33
Basuki, Achmad	37
Chung, Soon	14, 15
Dhandapani, Sowmya	7
Elfahmi, Muhammad Harits Shalahuddin	20
Elsen, Rickard	33
Fariza, Arna	37, 36
Ferdiana, Ridi	16
Galinium, Maulahikmah	8
Gunadi, I Gede	19
Hadiwijaya, Ryan	2
Hafidhoh, Nisa'ul	40
Harjoko, Agus	19, 21, 29
Hasim, Jauari	37, 36
Hendradjaya, Bayu	4, 5, 11, 22, 44
Hendriawan Nur, Wawan	3
Heryawan, Lukman	9
Indriasari, Theresia Devi	6
Irvan, Irvan	6
Irwan, Irwan	11
Kurniawan, Kabul	38
Laksmiwati, Hira	33, 41
Liem, Inggriani	2, 10, 40
Mahendra, Adhiguna	8

Mardiyanto, Sukrisno	5
Megatroika, Alvin	8
Munandar, Devi	28
Nata, Gusti Ngurah	26
Natali, Vania	10
Nattassha, Ruth	10
Nugroho, Hanung	35
Nugroho, Lukito	1
Nurhudatiana, Arfika	13
Ordiyasa, I Wayan	25
Perwitasari, Anggi	39
Praptini, Rina	4
Prasetya, Riza	37
Pratama, Bayu Yudha	31
Purnomo, Yohanes Sigit	6
Purnomosidi, Bambang	1
Putra, Agfianto	21, 29
Putri, Alifa	34
Rahadianto, Haris	36
Rahmawati, Dini	18
Ramdhani, Neila	16, 19
Rhaman, Md. Khalilur	23
Riasetiawan, Mardhani	17
Rusdah, Rusdah	32
Ruseno, Neno	8
Sabrina, Puspita	27
Santosa, Paulus	1
Santosa, Faulus Santoso, Hadi	29
Santoso, madi Saptawati, Putri	18, 20, 24, 26, 27, 39
Saputra, Kevin	43
Sarno, Riyanarto	31
Sevani, Nina	30
Solichin, Achmad	21
Sudarmilah, Endah	16
Sunindyo, Wikan	22, 34
Sumidyo, wikan Susanto, Adhi	16
Tantra, Junno	44
Wagey, Billy Charles	5
Wagey, Biny chanes Wardoyo, Retantyo	19, 32
wardoyo, iteaniyo	17, 54

Wibirama, Sunu	35
Widyani, Yani	18, 33, 41, 44
Widyantoro, Dwi	43
Widyawan, Widy	1
Wijayanto, Titis	35
Winadi, Mumtaz	35
Winarko, Edi	32

# The Clustering of High Schools Based on National and School Examinations

A Case Study at Daerah Istimewa Yogyakarta Province

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Abstract— The purpose of Indonesian National Examination for High School Students is to measure and assess students' knowledge and competence in particular subjects. The result is also going to be used as one of consideration for mapping Indonesia's national education quality. Aside from National Examination (NE), each school also conducts School Examination (SE). Both examinations are supposed to represent quality of education since the examinations measure the competence of the same students. However, the results of both examinations are not always linear [1]. In fact, the need of NE in Indonesian education is still being pro and cons among society.

In order to identify whether NE and SE could be used to represent the quality of educations in Daerah Istimewa Yogyakarta Province, this paper describes the analysis of NE and SE score by performing data mining technique using Fuzzy C-Means clustering algorithm towards NE score and SE score independently. Furthermore, the clusters were analyzed using univariate Anova, Spearman correlation, and crosstabulation. The data used in this research are NE and SE scores of Natural Science Department and Social Science Department of all high schools in Daerah Istimewa Yogyakarta Province from academic year 2011/2012 to 2014/2015.

The results of cluster analysis are three different clusters of NE in Natural Science Department, three different clusters of NE in Social Science Department, three different clusters of SE in Natural Science Department, and three different clusters of SE in Social Science Department for each year. The clusters are significantly separated. There is an opposite direction relationship between clusters of NE and SE. The relationship is weak which means there is no guarantee that a school which belongs to cluster-i of NE will be in the cluster-i of SE. Both for NE and SE memberships, only few members migrated from one cluster to another across years. The number of schools having the same cluster of NE and SE in each department varies from year to year, but generally less than 22%. The migrations of NE and SE cluster members from higher cluster to lower one and vice versa also vary.

In addition, there is inconsistency clustering based on NE and SE. Since SE is not standardized and indeed is a formative test, there might be subjective aspects involved in grading the students.

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Therefore, if the government intends to map Indonesia's national education quality, national examination is more suitable than school examination for this purpose.

Keywords— knowledge discovery in databases; Indonesian National Examination; Clustering; fuzzy c-means; Anova; Spearman correlation; crosstab

#### I. INTRODUUTION

Several efforts in the area of curriculum, quality and professionalism of teachers, infrastructure, and evaluation systems have been done by Indonesian government to improve the quality of education in Indonesia. In term of evaluation, Indonesian government has established a government law (PP no. 19/2005 article 63) containing several types of evaluation in education, namely evaluation by teachers, evaluation by schools, and evaluation by government as well.

Evaluation by teachers is a formative evaluation during the learning process in the whole semesters. Evaluation by schools are usually performed at the end of each semester and usually referred as School Examination (SE). Evaluation by government, which is usually referred as National Examination (NE), is performed at the end of the period of study in particular levels (elementary school, secondary school, and high school). Exam questions in NE are standardized by government, while exam questions in SE are composed by each school. In addition, each school has an authority to determine minimum completeness criteria for their students to pass the exams.

Based on PP no. 19/2005 article 63, the purpose of Indonesian National Examination for High School Students is to measure and assess students' knowledge and competence in particular subjects. The result is also going to be used as one of consideration for mapping Indonesia's national education quality.

In the year of 2010, the result of NE was the only factor to determine whether students should pass from a particular level of education or not. In this period, there were 267 schools whose students 100% failed in NE (Kompas, 28/4/2010). During

academic year 2011/2012 until 2014/2015, both NE and SE scores contributed to the Final Score (FS) that was used to determine students' graduation from High School.

Both NE and SE are supposed to represent quality of education since the examinations measure the competence of the same students. However, the authors found that the results of both examinations are not always linear [1]. In fact, the need of NE in Indonesian education is still being pro and cons among society. Therefore, the authors conduct a further research to identify whether NE and SE could be used to represent the quality of educations by taking a case study at Daerah Istimewa Yogyakarta Province.

Several studies about NE in particular places in Indonesia have been performed as can be found in [1], [2], [3], and [4]. Educational data mining using clustering methods have also been used by [5], [6],[7],[8],[9] and [10] for mapping the quality of schools based on the results of NE.

Other algorithms that are also frequently used is Fuzzy Clustering Means (FCM). FCM algorithm was first proposed by Ruspini [11] and later updated by Dunn [12] and Bezdek [13]. FCM clustering is a technique in which the existence of each data point in a cluster is determined by the degree of membership. The purpose of the Fuzzy C-Mean algorithm is to find the centroid by minimizing the objective function. FCM has been used for geo statistical data analysis problems in [13] while Lu et.al. in [14] used FCM for meteorological data.

In the domain of education in Indonesia, research using Fuzzy C-Means algorithm had been done by Luthfi [15] who clustered lecturers' teaching performance. Fuzzy C-Means for distribution of lecture participants is carried out by Susanto [16]. Karti in [17] used C-Means algorithm and Fuzzy C-Means clustering to cluster cities in East Java Province based on the similarity of high schools and vocational schools education indicators, namely the Net Enrolment Ratio (NER) and the Gross Enrolment Ratio (GER).

However, the researchers have not found researches that study NE in comparison with SE. The map of school quality based on NE and SE, the characteristics of each clusters as well as the consistency of each cluster across years have not been found yet. As the following part of the first research in [1], in this paper the researchers describe the analysis of NE score and SE score by performing data mining technique using Fuzzy C-Means clustering algorithm towards NE score and SE score independently. The clusters were then analyzed using univariate ANOVA, Spearman correlation, and crosstab. Univariate Anova was used to validate whether the clusters are significantly separated. Spearman correlation was used to identify the relationship between NE cluster memberships and SE cluster memberships. In addition, Spearman correlation was also used to identify the membership of each cluster across several years as well. Crosstabulation was used to find the percentage of schools whose memberships in NE clusters are the same with the memberships in SE clusterss.

The data used in this research are NE and SE scores of all high schools in Yogyakarta Province from academic year 2011/2012 to 2014/2015.

The result of this research hopefully could be used to identify whether NE and SE could be used to represent the quality of educations in Daerah Istimewa Yogyakarta Province.

#### II. RESEARCH METHODOLOGY

This research uses Knowledge Discovery in Databases [18] process. The implementation of the methodology is as follows:

- Selection process as the process to select target data. The 1. data available from the website of research and development division of The Indonesian Ministry of Education and Culture. The data can be downloaded in the form of spreadheet. It contains the result of SE for primary schools, secondary schools, and high schools in all provinces in Indonesia from year 2011 until 2014 and the result of NE as well as FS from year 2010 until 2014. For this research, the data selected as target data are high schools' SE and NE from year 2011 until 2014. The data from year 2010 is not included due to the fact that it does not contain SE since NE is the only factor to determine the students graduation during year 2010. The data of NE and SE are limited only for Natural Science Department and Social Science Department of all high schools (state and private schools) in Daerah Istimewa Yogyakarta Province. The province was selected as the case study since the province is the residence of the researchers so that it is more feasible to realize the plan to conduct in depth analysis in the future. Language Department was excluded as target data since only several schools have language department. The data of Natural Science Department contains fifteen columns as described in Table I, while the data of Social Science Department contains almost similar colums as described in Table II.
- 2. Preprocessing process to clean noise in the data. Theoretically, in this step the data should be cleaned from noises. However, the data that were downloaded from official website of The Indonesian Ministry of Education and Culture did not contain any noise. Therefore, there was no data cleaning performed. However, it was identified that the number of schools from year 2011 to 2014 were not the same due to the facts that there were several new schools in particular years or there were no students in several schools in particular years. These facts would not affect analysis, so that there is no need to perform data cleaning.
- 3. Transformation process to convert data from the official website into particular form that is ready to be mined. Since the data were clustered independently for each year and each type of exam (NE/SE) for each department, there was no transformation needed to convert data as described in Table I and Table II. Therefore, there were 16 dataset in the form of spreadsheet as described in Table III.

TABLE I. DATA DESCRIPTION OF NATURAL SCIENCE DEPARTMENT

No	Field
1	Number
2	School code
3	School name
4	School status (state/private)
5	Number of students joining examination
6	Number of students failed
7	The percentage of students failed
8	Mean score of Indonesian language test
9	Mean score English test
10	Mean score Mathematics test
11	Mean score Physics test
12	Mean score Chemical test
13	Mean score Biology test
14	Total of mean score of six courses being tested in Natural Science Department
15	The rank of school

TABLE II. DATA DESCRIPTION OF SOCIAL SCIENCE DEPARTMENT

No	Field
1	Number
2	School code
3	School name
4	School status (state/private)
5	Number of students joining examination
6	Number of students failed
7	The percentage of students failed
8	Mean score of Indonesian language test
9	Mean score English test
10	Mean score Mathematics test
11	Mean score Ecoomics test
12	Mean score Sociology test
13	Mean score Geography test
14	Total of mean score of six courses being tested in Social Science Department
15	The rank of school

- Data mining process by performing data mining technique 4 using Fuzzy C-Means clustering algorithm towards NE score and SE score independently. The algorithm was implemented using Java programming language. The program read 16 datasets that have been prepared in step 3, one file at a time. In each running, the program was executed using the following parameters: number of clusters = 3, weighting exponent = 2, maximum number of iterations = 100, and error tolerance = 0.001. Since the research would identify the consistencies of clustering across years, in each running the program read each dataset in Table III independently. The results of the program are member of each clusters along with the centroid of each clusters.
- 5. Evaluation to analyze the clusters formed in step 4 were then performed by using several analytical tools as follow:
  - a. Univariate Anova to validate that the clusters are significantly separated.
  - b. Spearman correlation to identify the relationship between NE cluster memberships and SE cluster memberships. In addition, Spearman correlation is also used to identify the membership of each cluster across several years as well.
  - c. Crosstabulation to find the percentage of schools whose memberships in NE cluster as well as SE cluster remain the same.
- 6. Interpretation was done to describe the meaning of the results of the above analytical tools so that it can be easily understood by common people.

#### **III. RESULTS AND DISCUSSIONS**

#### A. Fuzzy C-Means Clustering

In this research, clustering was performed to cluster schools based on NE and SE scores. The clustering will classify schools such that schools with similarity will be in the same cluster. The clusters will have high internal homogeneity and high external heterogeneity as well. To be more effective, cluster analysis was preceded by outlier detection using histogram. Based on the histograms, there was no outlier founded in the data of NE and SE scores for both natural science and social science departments.

By using Fuzzy C-Means, schools were classified into 3 groups according to the categorization of schools determined by The Ministry of Education and Culture, namely group A, B, and C that represent schools with high score category, middle score category, and low score category respectively. Table IV to Table VII describe the centroid of each cluster for each year. Cluster 1 represents schools with the highest score, cluster 2 represents schools with middle score, while cluster 3 represents the lowest one. From the table, it can be identified that the number of schools in each cluster varies. Several schools migrate from cluster 1 to 2, 2 to 3, and vice versa. The migration of schools can be identified from the membership of each cluster.

No	Name of dataset	Type of Exam	Department	Year
1	NE_ND_11	National Exam	Natural Science	2011
2	NE_ND_12	National Exam	Natural Science	2012
3	NE_ND_13	National Exam	Natural Science	2013
4	NE_ND_14	National Exam	Natural Science	2014
5	NE_SD_11	National Exam	Social Science	2011
6	NE_SD_12	National Exam	Social Science	2012
7	NE_SD_13	National Exam	Social Science	2013
8	NE_SD_14	National Exam	Social Science	2014
9	SE_ND_11	School Exam	Natural Science	2011
10	SE_ND_12	School Exam	Natural Science	2012
11	SE_ND_13	School Exam	Natural Science	2013
12	SE_ND_14	School Exam	Natural Science	2014
13	SE_SD_11	School Exam	Social Science	2011
14	SE_SD_12	School Exam	Social Science	2012
15	SE_SD_13	School Exam	Social Science	2013
16	SE_SD_14	School Exam	Social Science	2014

TABLE III. DATASET OF NE AND SE

#### TABLE V. NE CLUSTER CENTROID OF SOCIAL SCIENCE DEPARTMENT

	Cluster		
	1	2	3
Year 2011	7.61	6.32	4.85
Number of schools	65	65	34
Year 2012	7.54	6.21	4.90
Number of schools	52	64	47
Year 2013	7.76	6.19	5.03
Number of schools	31	61	70
Year 2014	7.45	5.82	4.57
Number of schools	35	55	69

TABLE VI. SE CLUSTER CENTROID OF NATURAL SCIENCE DEPARTMENT

		Cluster	
	1	2	3
Year 2011	8.86	8.36	7.89
Number of schools	45	62	27
Year 2012	9.03	8.45	7.99
Number of schools	38	49	48
Year 2013	9.16	8.58	8.11
Number of schools	28	56	53
Year 2014	9.17	8.55	8.12
Number of schools	19	59	63

 TABLE IV.
 NE CLUSTER CENTROID OF NATURAL SCIENCE DEPARTMENT

		Cluster	
	1	2	3
Year 2011	7.72	6.61	5.12
Number of schools	52	50	32
Year 2012	7.58	6.40	5.00
Number of schools	54	53	28
Year 2013	7.72	6.30	5.25
Number of schools	32	59	46
Year 2014	7.20	5.76	4.58
Number of schools	34	51	56

TABLE VII. SE CLUSTER CENTROID OF SOCIAL SCIENCE DEPARTMENT

		Cluster	
	1	2	3
Year 2011	8.78	8.35	7.80
Number of schools	43	86	35
Year 2012	8.91	8.45	8.03
Number of schools	47	66	50
Year 2013	9.04	8.54	8.13
Number of schools	37	73	52
Year 2014	9.21	8.56	8.18
Number of schools	29	65	65

#### B. Univariat Anova

To make sure that the clustering based on NE and SE result on clusters which is significantly independent, Univariat Anova was performed on each academic year by checking whether there are differences of NE and SE mean scores between clusters.

Significant differences indicate that clustering is able to significantly differentiate schools into 3 groups.

Based on Anova tables, it can be concluded that the clustering of schools based on the mean score of NE and SE for all academic years have resulted on three significantly different clusters. Cluster 1, 2, 3 consecutively are schools with the highest mean score of NE or SE (first order), second order, and third order. As an illustration, Table VIII is an example of Anova table for NE score in year 2011. Significance value 0.000 (which is less than  $\alpha = 0.05$ ) shows that the clustering has resulted on significantly different clusters. All Anova tables for NE score and SE score in year 2011 to year 2014 has significance value 0.000.

#### C. Spearman Correlation

The using of Spearman Correlation is intended to identify the relationship between membership of NE and SE clusters. High correlation between the two shows there is a consistency of clusters based on NE and SE. Ideally, a school that is categorized in cluster 1 of NE will be categorized as cluster 1 of SE as well.

The result of Spearman correlation of NE and SE membership for both natural science and social science department are described in Table IX and X. Based on those tables, negative correlation indicates that there is an opposite relationship between clusters based on NE and SE. The correlation is weak (the absolute values is less than 0.5) which means there is no guarantee that a school which belongs to cluster-i of NE will be in the cluster-i of SE. Negative and weak correlation also shows that several schools which are belong to high NE score cluster will tend to be members of lower SE score cluster. Few schools in particular SE cluster migrate into lower NE cluster.

Using Spearman correlation it can also be identified the cluster membership correlation across years. Higher correlation indicates that schools will tend to be in the same cluster. For NE, the membership correlation which is between 0.581 to 0.855 (for natural science department) and 0.670 to 0.868 (for social science department) shows that only few members migrated from one cluster to another across years. The conclusion also applies for SE as represented by the membership correlation which is between 0.541 to 0.758 (for natural science department) and 0.503 to 0.800 (for social science department).

#### D. Crosstabulation

The using of crosstab is to support the Spearman correlation. From crosstab it can be identified the percentage of schools whose memberships in NE clusters are the same with the membership in SE clusters. It can also be counted the percentage of schools whose memberships change. Table XI shows an example of crosstabulation between NE and SE clusters in year 2011. Crosstab for other years are not presented in this paper

TABLE VIII. ANOVA TABLE OF NE IN YEAR 2011

	Sum of Squares	df	Mean Square	F	Sig.
Between Groups	130.916	2	65.458	453.030	.000
Within Groups	18.828	131	.144		
Total	149.845	133			

TABLE IX.	SPEARMAN CORRELATION BETWEEN NE AND SE OF
	NATURAL SCIENCE DEPARTMENT

	Cluster_SE_ 11	Cluster_SE_ 12	Cluster_SE_ 13	Cluster_SE_ 14
Cluster_NE_11	-0.307			
Cluster_NE_12		-0.261		
Cluster_NE_13			-0.400	
Cluster_NE_14				-0.376

TABLE X.	SPEARMAN CORRELATION BETWEEN NE AND SE OF
	SOCIAL SCIENCE DEPARTMENT

	Cluster_SE_ 11	Cluster_SE_ 12	Cluster_SE_ 13	Cluster_SE_ 14
Cluster_NE_11	-0.217			
Cluster_NE_12		-0.275		
Cluster_NE_13			-0.412	
Cluster_NE_14				-0.454

TABLE XI. CROSSTABULATION OF NE CLUSTERS AND SE CLUSTERS OF NATURAL SCIENCE DEPARTMENT IN YEAR 2011

			Cluster_SE11			
			1	2	3	Total
Cluster_NE11	1	Count	13	20	19	52
		% of Total	9.7%	14.9%	14.2%	38.8%
	2	Count	15	29	6	50
		% of Total	11.2%	21.6%	4.5%	37.3%
	3	Count	17	13	2	32
		% of Total	12.7%	9.7%	1.5%	23.9%
Total		Count	45	62	27	134
		% of Total	33.6%	46.3%	20.1%	100.0%

Based on crosstabulations, it can be recognized several facts as follows:

- 1. The number of schools having the same cluster of NE and SE in each department varies from year to year, but generally less than 22%.
- 2. The migrations of NE and SE cluster members from higher cluster to lower one and vice versa also vary.

The above facts emphasize that there is inconsistency clustering based on NE and SE. As elaborated in [1], the clustering of NE will tend to end up in clusters with wide interval of NE score, while clustering of SE will end up in clusters with narrow interval and high score of SE.

Since SE is not standardized, there might be subjective aspects involved in grading the students. Since SE indeed is a formative test, schools would try their best to help students passing minimum completeness criteria and finally combine with NE score that will contribute to the final score of student's graduation. Interview with several educational practitioners revealed these facts.

Therefore, if the government intends to map Indonesia's national education quality, national examination is more suitable form of test for this purpose.

#### **IV. CONCLUSIONS**

The results of cluster analysis are three different clusters of NE in Natural Science Department, three different clusters of NE in Social Science Department, three different clusters of SE in Natural Science Department, and three different clusters of SE in Social Science Department for each year. The clusters are significantly separated. There is an opposite direction relationship between clusters of NE and SE. The relationship is weak which means there is no guarantee that a school which belongs to cluster-i of NE will be in the cluster-i of SE. Therefore, it should be carefully examined which is actually represent the quality of education, either NE score or SE score.

Both for NE and SE memberships in all departments, only few members migrated from one cluster to another across years. The number of schools having the same cluster of NE and SE in each department varies from year to year, but generally less than 22%. The migrations of NE and SE cluster members from higher cluster to lower one and vice versa also vary.

Future works will be performed to analyze clusters toward each subject in NE to identify the possible strengths and weaknesses of each school. Analysis of new data from the year 2015 in which NE are not used as component to determine student's graduation will also be performed to study the effect of the new policy towards students and school achievement.

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#### REFERENCES

- A. Dwiatmoko, P. Rosa, and R. Gunawan, "Analisis Statistis Data Nilai Ujian Nasional dan Nilai Ujian Sekolah Menengah Atas di Propinsi Daerah Istimewa Yogyakarta," *J. Ilm. Media Tek.*, vol. 14, no. 2, pp. 89– 95, 2015.
- [2] Rasto, Nahadi, and S. A. Soesilawaty, "Analisis Peta Kompetensi Hasil Ujian Nasional dan Model Pengembangan Mutu Sekolah SMA di Jawa Barat (Survey di Kabupaten Garut dan Kabupaten Tasikmalaya)," Artik. Has. Penelit., vol. 2011, pp. 1–17, 2011.
- [3] A. Jaedun and E. Listiyani, "Pemetaan Mutu Pendidikan di SMA Kabupaten Cilacap dan Banyumas Berdasarkan Analisis Hasil Ujian Nasional," in *Prosiding Seminar Nasional*, 2012.
- [4] Zulkardi, "Model Peningkatan Mutu Pendidikan SMA di Kota Prabumulih, Kabupaten Ogan Ilir, dan Kabupaten Ogan Komering Ilir," in Prosiding Seminar Nasional Hasil Penelitian Universitas Sriwijaya, 2011.
- [5] R. Khairunnisa and H. Fajriya, "Pengelompokan Sekolah Dasar Berdasarkan Rata-Rata Nilai Ujian Akhir Sekolah Berstandar Nasional Di Setiap SD/MI di Kota Yogyakarta Dengan Menggunakan Metode Hierrarkhi Clustering," in *Naskah Seminar Kerja Praktek, FMIPA, UII* Yogyakarta, 2013.
- [6] Pakpahan and M. Juni, "Pengelompokan Sekolah Menengah Pertama Berdasarkan Rata-Rata Nilai Ujian Akhir Nasional Di Kota Binjai dengan Analisis Hierrarkhi Clustering," in *Naskah Skripsi Universitas Negeri Medan*, 2013.
- [7] Prihatiningtyas, "Analisis Hasil Ujian Nasional Tingkat SMA di Kabupaten Banyumas Menggunakan Analisis Cluster dan Biplot," in Skripsi Program Studi Statistika, Universitas Diponegoro, Semarang, 2011.
- [8] Habibi and A. Ahmad, "Pemetaan UN Propinsi di Indonesia Melalui Analisis Klaster Penggabungan Pautan Tunggal," in *Thesis JBPTITBPP*, *ITB Bandung, Indonesia*, 2009.
- [9] Fadhli, "Analisis Kluster Untuk Pemetaan Mutu Pendidikan di Aceh," in Thesis Jurusan Matematika Universitas Gadjah Mada Yogyakarta, 2011.
- [10] Sjafrudin and Asep, "Analisis Hasil Ujian Nasional Madrasah Tsanawiyah Tahun 2008," 2009.
- [11] E. H. Ruspini, "A new approach to clustering," *Inf. Control*, vol. 15, no. 1, pp. 22–32, 1969.
- [12] J. C. Dunn, "A Fuzzy Relative of the ISODATA Process and Its Use in Detecting Compact Well-Separated Clusters," J. Cybern., vol. 3, no. 3, pp. 32–57, 1973.
- [13] J. C. Bezdek, R. Ehrlich, and W. Full, "FCM: The fuzzy c-means clustering algorithm," *Comput. Geosci.*, vol. 10, no. 2–3, pp. 191–203, 1984.
- [14] Y. Lu, T. Ma, C. Yin, X. Xie, W. Tian, and S. Zhong, "Implementation of the Fuzzy C-Means Clustering Algorithm in Meteorological Data," *Int. J. Database Theory Appl.*, vol. 6, no. 6, pp. 1–18, 2013.
- [15] E. T. Luthfi, "Fuzzy C-Means Untuk Clustering Data (Studi Kasus: Performance Mengajar Dosen)," Semin. Nas. Teknol. 2007 (SNT 2007), vol. 2007, no. November, pp. 1–7, 2007.
- [16] S. Susanto and Ernawati, "Pembagian Kelas Peserta Kuliah Berdasarkan Fuzzy Clustering dan Partition Coefficient and Exponential Separation ( PCAES) Index," *J. Teknol. Ind.*, vol. X1 (2), no. APRIL 2007, pp. 143– 154, 2007.
- [17] H. S. Karti, "Pengelompokan Kabupaten / Kota di Provinsi SMA / SMK / MA dengan Metode C-Means dan Fuzzy C-Means," J. Sains dan Seni Pomits, vol. 2, no. 2, p. D–288, 2013.
- [18] P.-N. Tan, M. Steinbach, and V. Kumar, *Introduction to Data Mining*, First. Boston, MA, USA: Addison-Wesley Longman Publishing, 2005.