Analyses of Student Questionnaires for Faculty Developments

Shigeichi Hirasawa *

Abstract—Combining statistical analyses and information retrieval techniques, an efficient way for knowledge discovery from questionnaires is discussed. Since usual questionnaires include questions answered by a fixed format and those by a free format, it is important to introduce the viewpoints of both data mining and text mining, where the answers by the fixed format is called "items", and those by the free format, simply "texts". In this paper, an algorithm for processing answers with both the items and the texts is presented. An algorithm for extracting important sentences from documents is also presented. Using these algorithms combined with statistical techniques, a method for analyzing the questionnaires is established. The method is applied to a case of improvements for the quality of education for which the student questionnaire is executed to a class, and we obtain useful knowledge which leads to faculty developments.

Keywords—questionnaire, information retrieval, PLSI model, classification, clustering, data mining, text mining, statistical analysis, faculty development, improvement of education

1 Introduction

Universities are expected to develop useful and effective programs for classroom management and to improve the quality of education at all times. To perform these activities, student questionnaires are often used. By establishing a class model for an individual class[GSIIH03],[IIGSH03-b], we have evaluated student characteristics, the degree of satisfaction and final scores, and their relationships for a set of students or subsets of them[IIGSH03-b]. The present author has collected data from student questionnaires for the past four years, the class considered here is mainly "Introduction to Computer Science", in the second academic year, Department of Industrial and Management Systems Engineering, School of Science and Engineering, Waseda University.

First, we present techniques for (1) classification or clustering for documents with fixed format and free format [IIGSH03][HC03][HC04], and (2) extraction of important sentences or important words from documents [IIGH02][SIGIH03][ISH03]which enable us to briefly understand the contents of the documents. Using the traditional statistical techniques, (3) interpretation of characteristics of the set of documents. Next, we establish a class model used for the class of " Introduction to Computer Engineering ". The model has inputs information of students such as implicit characteristics, a prior knowledge, interested areas, intention, etc. and has output such as final score and degree of satisfaction given by the students. Finally, we apply the above techniques into student questionnaire analyses.

Problems of partitioning students of the class into a few subclasses by their characteristics are evaluated. The purposes of these problems are to improve the degree of satisfaction of the students and increase the effectiveness of education.

In this paper, we show a questionnaire analysis model in section 2. The student questionnaire is mainly discussed in section 3. Section 4 discusses results of analyses. Concluding remarks are given in Section 5.

2 Questionnaire Analysis Model

The method for analyzing the questionnaire is shown in Fig. 2.1 as a questionnaire analysis model.

First, a model for the object for which a questionnaire will be examined is generated. For example, we shall show a class model in this paper.

Second, a questionnaire is designed based on this model, which includes both the items and the texts as the answers. We call them documents. The number of the documents equals that of examinees , i.e., students in this paper.

Next, analyses are executed as follows:

- The set of documents is classified or clustered by the proposed algorithm for classification or clustering [HC03] [HC04] [HIAGS05] [IIGSH03a]. Note that both the items and the texts are processed together, not separately.
- (2) For the texts only, important sentences, or the parts of them are extracted from the documents by the proposed algorithm for extracting important information [IIGH02] [SIGIH03] [ISH03] [SIGH04]. These results are helpful to easily understand the opinions and directly give useful information of the classes or clusters.
- (3) For the items ¹ only, statistical techniques such as multiple linear regression analysis, discriminant analysis and so on, are used to analyze the characteristics of each set of members. If the amount of the data is extremely large, a data mining technique is also used to analyze them.

In (1), we have proposed the algorithm based on the probabilistic latent semantic indexing (PLSI) model [Hofmann99] [CH01] which is known to be one of the most powerful model in information retrieval systems. The proposed algorithm based on PLSI model exhibits good performance in classification or clustering especially for a small size of the document set. In (2), we

^{*} Department of Industrial and Management Systems Engineering, School of Science and Engineering, Waseda University, 3-4-1, Ohkubo, Shinjuku, Tokyo 169-8555 Japan. Phone:+81-3-5286-3290. Fax:+81-3-5273-7215. E-mail:hirasawa@hirasa. mgmt.waseda.ac.jp

¹ Information investigated attribute of the categorical data, e.g., the scores of examinations for students, is added to a sort of the items.

have also proposed the algorithm to select important sentences by extracting representative words or sentences based on Japanese language processing.

The results obtained by combining (1) and (3) give the profile of each class or cluster by the characteristics of the members. Combining (2) and (3) is also used for understanding the characteristics of the members of each class of cluster and these results give us useful information to manage the mass or improve the conventional systems.

Finally, actions are made based on the analyzed results. The actions are evaluated from the standpoint of their effectiveness, and a new model for the objective is generated by the feedback loop if necessary.



Figure 2.1: Questionnaire analysis model

3 Student Questionnaire

A class model in this object is shown in the Fig.3.1. A technique to extract requests of the students from the questionnaire is discussed by applying the questionnaire analysis model. First, relationships between the degree of satisfaction, score and the characteristics of the students and so on are presented as a class model. Next, the questionnaire is designed to verify the hypothesis given by this class model. Finally, according to the results of this questionnaire analysis together with the score of each student, we evaluate the degree of satisfaction, that of achievement in learning and characteristics of students. This knowledge is useful to manage the class. In many universities in Japan, recently attentions to the quality assurance of the education program by Japan Accreditation Board for Engineering Education (JABEE) becomes important as for improvements of a class.

3.1 Class model

We have proposed a class model for the class "Introduction to Computer Science" as shown in Fig. 3.1 [IIGSH03-b].

The implicit characteristics of each student are essentially measured by questionnaire. While explicit characteristics are objectively given by numerical data of the class. These characteristics generate explanatory variables. Then each student yields his or her



Figure 3.1: Class model

final score and the degree of satisfaction as the result of the class. The degree of satisfaction is also measured by questionnaire. The final score and the degree of satisfaction play as criterion variables in this model. Besides these variables, we can expect to get some information regarding to classroom management such as partition of the class. Usually there exist many differences between each student in level, interested area, experiences, and motivation before beginning the class, since the class is opened in Department of Industrial and Management Systems Engineering. Hence a proper partition of the class depending on features such as the future plan of each student is desirable. The partitions shown in Table 1 can be considered, where G stands for a generalist course, S, for a specialist course by estimating his or her future work. According to this model, we can effectually design the questionnaire. In Fig. 3.1, note that explicit characteristics can act as both input variables and output variables.

3.2 Design of Questionnaire

A questionnaire was applied to the class: "Introduction to computer science". It consists of the initial questionnaire (IQ) and the final questionnaire (FQ). Scores of technical report (TR) submitted every week, and those of the midterm test (MT) and final test (FT) are explicit characteristics of each student. We analyze them by using statistics, data mining, and information retrieval techniques that include classification and clustering. The example of contents of a questionnaire is shown in Table 3.1 and in Table 3.2. The time schedule for class is depicted in Fig. 3.2.

4 Questionnaire analyses

4.1 Verification of class model by IQ

Before starting the class, we discuss problems on the class management and the lecture plan. By using only IQ, the following problems are considered:

Table 3.1: Data of class

Exercise	Contents
Initial Questionnaire (IQ) Item type Text type	7 questions (4-20 sub-questions each) 5 questions (250-300 characters in Japanese each)
Midterm Test(MT)	5 subjects
Technical Reports (TR)	11 times (each 1-2 subjects)
Final Test (FT)	5 questions
Final Questionnaire (FQ)	
Item type	6 questions (6-21 sub-questions each)
Text type	5 questions (250-300 characters in Japanese each)

Table 3.2: Contents of questionnaire

Eventice		Evamples (sub questions)
IQ	Item- type	
	Text- type	 Write about your knowledge and experience on computer. What kind of work will you have after graduation? What do you imagine from the name of this dass subject name?
Б	recise	Examples (sub questions)
FQ	Item- type	Could you understand the contents of this lecture? Was the midterm test difficult? Was it easy to read the handwritings on the white-board? Do you think the contents of this lecture to be useful to yourself? Do you want to finish this course even if it is optional? Which are you interested in applied technology or the fundamentals of computers? Which do you choose class (S) or class (G)?
	Text- type	 Do you want to be a member of laboratories related to the information technology? In the future, will you get a job in industries related to the information technology? Did your image on computers change after taking this lecture?

This questionnaire is made in WEB form, and it is on the following Web Site. http://hirasa.mgmt.waseda.ac.jp/users/comp-eng/

- (1) Prediction of the scores
- (2) Partition of the students of the class
- (1) Prediction of scores

For explaining the score by only IQ, it is difficult to distinguish even whether he get a high score or not (The rate of contribution of the multiple linear regression analysis was 51%.)

(2) Partition of students of class

The purpose of the partition of students is to improve the effect of education by adequately partitioning the students of the class based on their interested areas, levels, or intentions. Since the partition is made at the beginning of the class, we must make it by IQ only.

We have the following three partitions. As shown in Table 4.1, we provide two courses depending on the contents of topics of the lecture:

(a) Partition by the contents of topics

Class S (specialist): technical and professional topics Class G (generalist): wide and shallow technical topics

We can also provide two classes depending on the level of the lecture and on the management of the class:



Figure 3.2: Time schedule for class

Table 4.1: Contents of topics

Class	Contents
	- History of computers, fundamental con-
	cepts in computer
	- Basics of architecture
Class G	- Basics of hardware
	- Basics of software
	- Applications of information technology
	etc.
	- Architecture(stack machine, binary sys-
	tem, processor architecture)
Class S	- Hardware(logic design, logical circuit, au-
	tomaton)
	- Software(operating system, UNIX, lan-
	guage processor)
	etc.

- (b) Partition by the student's level Class H: a higher level Class L: a lower level
- (c) Partition by the class managing method:Class E: exercise- and practice-based lectureClass T: test-based lecture

A partition (a) of Class S and Class G is examined by a clustering algorithm using both the item-type and the text-type questionnaire [HC03] of only IQ. Since this algorithm requires representative vectors (pseudo documents), they are obtained from the same questionnaire by graduate students (or the senior students whose jobs in future were decided). Then clusters are automatically generated.

The result of this clustering compared with student's own choice are shown in Table 4.2. The characteristics extracted by discriminant analysis are shown in Table 4.3.

The prediction error for classification is summarized in Table 4.4, where the prediction error is calculated

Table 4.2: Classification of Class G and Class S

ductoring	A student's own choice		
Clustering	G	S	Total
G	29	20	49
S	34	28	62
Tota	63	48	111

Table 4.3: Characteristics of Class G and Class S



by the difference between the automatic clustering and student's choice given by the item-type questionnaire of FQ.

Table 4.4: Prediction error for partition

(a) Sor G	(b) H or L	(c) E or T
0.40	0.40	0.33

4.2 Verification of class model by IQ and FQ

Let us try to explain (1) the scores, (2) the degree of satisfaction, and (3) the favorite partition of the students by the item-type questionnaire of IQ and FQ.

(1) Scores of students

We expect to explain the scores of the midterm test (MT) and of final test (FT)(as intermediate criterion variables) by the item-type questionnaire (as explanatory variables) of IQ and FQ. The result by the multiple linear regression analysis is shown in Table 4.5. Summarized sentences [IIGH02] extracted from the text-type questionnaire of IQ and FQ based on the scores are shown in Table 4.6.

(2) Degree of satisfaction

Similar to the above experiment, the item-type questionnaire (as explanatory variables) of IQ and FQ can explain the degree of satisfaction (as criterion variables) for the contents of topics and for the class management as shown in Table 4.7.

(3) Favorite partition

The reasons why the students choose Class G or S, and Class E or T as their favorite partition (as criterion variables) are shown in Table 4.8 and 4.9, respectively. Table 4.5: Explanation of scores by item-type questionnaire



Table 4.6: Summarized sentences extracted from texttype questionnaire

Score	Example of Sentences		
High Over 70	I think that this class is the one for giving interest to a computer. Since I was interested in the structure of a computer, I wants to participate the lecture eagerly. However, since I have almost no prior knowledge, I am worked in the ability to catch up with a class. Since I think that deep understanding in the field of information technology cannot be obtained without knowledge of computer structure, I want to this opportunity.		
Low Under 69	 I can only use a few functions of computers. For example, Internet, Excel or Word etc. I cannot effectively use a personal computer. Therefore, I think that I want to know various things related to computers. I cannot imagine the contents of the class from the name of subject "Introduction to computer science" well. And this subject would be uninterested for me. 		

4.3 Discussions

- (1) It is difficult to predict the student's final score by only IQ, and there is no explanation capability by the linear regression analysis. This can be, however, thought probably to be a natural result.
- (2) Although the prediction error rates of the partition problems are 30 40[%] for which only the item-type questionnaire of IQ are used, combining with the important sentences extracted from the text-type questionnaire gives useful information for managing the class. According to the characteristics of each class, we can improve the quality of education, and can use the result to guide students.
- (3) The student's own choice is insufficient for partition of Class G and Class S. Note that this problem is noto objective partition.
- (4) It is possible to explain the student's final score by IQ and FQ. However, it is difficult to get the suggestion to improve the student's score, although we can get a student's tendency.
- (5) It is a little difficult to explain the degree of satisfaction regarding the class management, but easy to explain that regarding the contents of topics by IQ and FQ.
- (6) It is possible to explain the favorite partition to the students by IQ and FQ. This suggest us a proper partition to the next year by taking into account causal relations obtained in this year.

Table 4.7: Explanation of degree of satisfaction by item-type questionnaire

Satisfaction with Contents of the dass

Explanatory variable	Partial regression coefficient	/- value	
FQ: I want to study this field after the class	1.5	5.6	
FQ: The reports are necessary for every week.	1.0	5.4	
FQ: I actively attended the class	0.9	4.2	
FQ: This lecture is necessary for our department.	0.8	4.1	
FQ: I had been interested in contents of a lecture.	0.9	3.9	
IQ : I want to have a qualification related to information technologies.	-1.0	-3.6	
IQ : Attendance should be taken	-0.5	-3.5	
IQ : How many days in a week in average do you come to university?	-1.3	-3.2	
IQ : I don't care if I lost a credit.	0.9	3.0	
IQ : I prefer science to literature.	0.4	2.8	
Satisfaction for the Class Management			
	Dartial		

Explanatory variable	regression coefficient	value
FQ: The reports are necessary for every week.	1.6	5.1
FQ: The degree of interest of the contents of this class.	0.2	4.1
IQ : I clearly have an object to learn in this class.	-1.4	-3.8
FQ: I am interested in the contents of this class.	1.5	3.7
IQ . I finished this class even if it is optioned.	1.3	3.3
IQ : This class is sufficient for a half year.	1.2	3.1
IQ : I don't care even if I lost a credit.	1.5	3.0
IQ : I checked a syllabus.	0.6	2.7
FQ: I finished this class even if it is optioned.	0.9	2.6
FQ: I want to have a qualification related to information technologies.	-0.9	-2.6
FQ: I can use the most of a PC by this class.	-0.9	2.2
FQ: I made the reports by myself.	0.8	-2.0

Contribution ratio= 0.60

Table 4.8: Explanation of favorite partition Class G or Class S

Explanatory variable	Partial regression coefficient	F -ratio
FQ: The degree of interest point of the contents of this lecture.	11.1	-0.2
IQ : I have never never assembled a PC.	7.2	-3.1
IQ : I am a woman,	6.7	2.1
IQ : How many years have you used your own PC.	6.6	-0.4
IQ : For how many years have you used a PC?	5.8	0.2
FQ: The degree of interest in areas related to the information.	5.4	0.1
FQ: This lecture is sufficient with in a half year.	4.8	-0.6
FQ: The degree of satisfaction of the contents of this lecture.	3.6	-0.2

Class S: specialist Class G: generalist Mis-distinction ratio 21. 1%

4.4 Conclusions and future works

It can be concluded that we obtain useful information to improve the class management by student questionnaire with both the item-type and the texttype. The result shows verification of the class model for "Introduction to computer science". The degree of satisfaction for the students should be investigated in detail as a future work. Questionnaire must be carried out to collect data for several years, and their time series analysis and the review of the model also remain as further studies.

4.5 Additional questionnaire

In 2005 spring, we introduced a subject registration system from the WEB which enable students for all schools of Waseda University to register their subjects from any place (all over the world) at any time. Prior to this introduction of the system, we examined it in trial for a few schools in 2004 spring. A questionnaire by students who joined the system, we obtained useful aspect to improve or to design the system performance criteria, the user interface, the system configuration, etc. The detailed are omitted here, the result of clus-

Table 4.9: Explanation of favorite partition Class E or Class T

Explanatory variable	Partial regression coefficient	F ratio
FQ: Attendance should be taken.	23.9	1.6
IQ: I am interested in the principle rather than the application of computers.	10.9	1.3
IQ : I want to study this field after the class.	10.4	-1.4
IQ : I learn a subject in which I am not interested.	8.5	1.0
FQ: This lecture is a necessary to me.	7.5	1.0
FQ: The report subject is better than the final test.	5.6	0.7
IQ : For how many years have you used the WEB?	5.2	-0.5
IQ : This lecture is necessary for our department.	5.0	-1.0
FQ: I want to have a qualification related to information technologies.	5.0	-0.9
Class E: exercise- and practice-based lecture Mis-distinct Class T: test-based lecture	ion ratio 11.	2%

tering is shown in Fig. 4.1 Ths figure shows a case of combination of two clusters and five clusters. We see that

- (1) The cluster $c_5(2)$ includes almost all students who used the system and answered " no problem " to use it. The rate of such positive students is approximately 0.93 in this cluster.
- (2) The clusters $c_2(1)$ and $c_5(2)$ include almost all students who used the system from their home (not from campus).

As a conclusion, the introduction of the system will be almost successful. A future investigation, however, would be necessary to check the contents of registration in detail.



Figure 4.1: Clustering of students for subject registration system

5 Concluding Remarks

We have shown the effective way for knowledge discovery from questionnaire by combining data mining and text mining techniques. One of the most remarkable points is to construct a questionnaire with both fixed format and free format and to simultaneously process them. The effective algorithm to extract the important sentences from questionnaire with free format (text) is also provided. A model which simply exhibits the real problem and the design of the questionnaire based on this model are important to successfully apply this method to actual problems. We have applied the method to improve the quality of education. Although there are many remained problems, we obtain effective information which leads to faculty developments [IGH05]. The developments of other algorithms such as data mining technique for classification and clustering which should be added to this method is a further research.

Acknowledgement

The author would like to thank Mr. T. Ishida, Mr. J. Itoh, Dr. T. Sakai, and Dr. M. Gotoh for their helpful and fruitful discussions.

References

- [BYRN99] R. Baeza-Yates and B. Ribeiro-Neto, Modern Information Retrieval, ACM Press, New York, 1999.
- [CH01] D. Cohn, and T. Hofmann, "The missing link - A probabilistic model of document content and hypertext connectivity," Advances in Neural Information Processing Systems (NIPS × 13), MIT Press 2001.
- [GIH03] M. Goto, T. Ishida, and S. Hirasawa, "Representation method for a set of documents from the viewpoint of Bayesian statistics," *Proc. IEEE* 2003 Int. Conf. on SMC, pp.4637-4642, Washington DC, Oct. 2003.
- [GITSH03] M. Gotoh, J. Itoh, T. Ishida, T. Sakai, and S. Hirasawa, "A method to analyze a set of documents based on Bayesian statistics," (in Japanese) Proc. of 2003 Fall Conference on Information Management, JASMIN, pp.28-31, Hakodate, Nov. 2003.
- [GSIIH03] M. Gotoh, T. Sakai, J. Itoh, T. Ishida, and S. Hirasawa, "Knowledge discovery from questionnaires with selecting and descripting answers," (in Japanese) *Proc. of PC Conference*, pp.43-46, Kagoshima, Aug. 2003.
- [HC03] S. Hirasawa, and W. W. Chu, "Knowledge acquisition from documents with both fixed and free formats," *Proc. IEEE 2003 Int. Conf. on SMC*, pp.4694-4699, Washington DC, Oct. 2003.
- [HC04] S. Hirasawa, and W. W. Chu, "Classification methods for documents with both free and fixed formats," *Proc. 2004 Int. Conf. Management Sciences and Dexision Making*, pp.427-444, Taipei, R.O.C., May 2004.
- [HIAGS05] S. Hirasawa, T. Ishida, H. Adachi, M. Goto, and T. Sakai, "A Document classification and its application to questionnaire analyses," (in Japanese) Proc. of 2005 Spring Conference in Information Management, JASMIN, pp.54-57, Tokyo, June 2005.
- [HIIGS03] S. Hirasawa, T. Ishida, J. Ito, M. Goto, and T. Sakai, "Analyses on student questionnaires with fixed and free formats," (in Japanese) *Proc.* of Comp. Edu. JUCE, pp.144-145, Sept. 2003.
- [Hofmann99] T. Hofmann, "Probabilistic latent semantic indexing," Proc. of SIGIR'99, ACM Press, pp.50-57, 1999.

- [IGH05] T. Ishida, M. Gotoh, and S. Hirasawa, "Analysys of student questionnaire in the lecture of computer science," (in Japanese) Computer Education, CIEC, vol.18, pp.152-159, July 2005.
- [IIGH02] J. Itoh, T. Ishida, M. Gotoh, and S. Hirasawa, "A method for extracting important sentences using co-occurrence similarities between words," (in Japanese) *IEICE 2002 FIT*, pp.83-84, Tokyo, Sept. 2002.
- [IIGSH03-a] J. Itoh, T. Ishida, M. Gotoh, T. Sakai, and S. Hirasawa, "Knowledge discovery in documents based on PLSI," (in Japanese) *IEICE 2003 FIT*, pp.83-84, Ebetsu, Sept. 2003.
- [IIGSH03-b] T. Ishida, J. Itoh, M. Gotoh, T. Sakai, and S. Hirasawa, "A model of class and its verification," (in Japanese) Proc. of 2003 Fall Conference on Information Management, JASMIN, pp.226-229, Hakodate, Nov. 2003.
- [ISH03] J.Itoh, T.Sakai, and S.Hirasawa, "A method for extracting parts of important sentences from Japanese documents using dependency trees," (in Japanese) IPSJ, *Tech. Rep. Natural language processing*, 158-4, pp.19-24, Nov. 2003.
- [Sakai99] T. Sakai, et al., "BMIR-J2: A Test Collection for Evaluation of Japanese Information Retrieval Systems," ACM SIGIR Forum, Vol.33, No.1, pp.13-17, 1999.
- [Salton71] G. Salton, The SMART Retrieval System -Experiments in Automatic Documents Processing, Prentice Hall Inc., Englewood Cliffs, NJ, 1971.
- [SIGIH03] T. Sakai, J. Itoh, M. Gotoh, T. Ishida, and S. Hirasawa, "Efficient analysis of student questionnaires using information retrieval techniques," (in Japanese) Proc. of 2003 Spring Conference on Information Management, JASMIN, pp.182-185, Tokyo, June 2003.
- [SIGH04] T. Sakai, T. Ishida, M. Gotoh, and S. Hirasawa, "A student questionnaires analysis system based on natural language expressions," (in Japanese) submitted to IEICE 2004 FIT.
- [Rissanen83] J. Rissanen, "A universal prior for integers and estimation by minimum description length," Ann. Statist. vol.11, no.22, pp416-431, 1983.