

K-Means clustering model to improve competency of diploma program graduates

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Abstract

Completion of studies by students is the result of the process carried out in the learning process in university. The problem that arises related to this condition is that many diploma students are often late in completing their studies or even dropping out of school. Departments at the university do not have a significant measuring tool in predicting student graduation rates based on the competencies mastered. This paper proposes a program model to predict the basic abilities of Information System Diploma Program students to be mapped in the level of ability in an effort to reduce the delay in the learning process. The method used in the program model was a quantitative method with a K-Means Clustering model approach. The results showed that there was a significant increase in the timely graduation rate for Information System Diploma Program.

Keywords: Graduates Competency, Diploma Program Student, Information System, K-Means Clustering, Higher Education;

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1. Introduction

Universities as entities in improving student competencies have an interest in developing learning programs. A structured learning program has the aim of delivering students to achieve competence and find the desired career (Collins et al., 2021). Education is a joint effort of universities and related stakeholders in creating a learning atmosphere and learning process so that students are actively able to develop their potential to have good self-control, personality, and character and skills needed by themselves and the community (Hidayah, 2021). Universities have an important role and position among the components of education. The purpose of higher education is to provide services that contain elements of limited norms but do not conflict with the nature of student development and can be well accepted by the community as the value of a good life (Moscardini et al., 2020; Nagarajan & Edwards, 2015). This is no exception to the Information System Diploma Study Program at Telkom University. Study programs that have many students with limited teaching resources must run an efficient and effective learning process. However, the study program is faced with fundamental problems that become obstacles in carrying out the learning process.

The educational process has a time limit determined by the university, according to the level of education. To take a level of education, students must follow the regulations that have been applied by the university. If exceeding the time limit, the student is declared to have failed in carrying out the educational process. Study time limit is the maximum time for a student to complete study in a higher education institution. The length of study taken is regulated or determined by universities in Indonesia (Pendidikan & Indonesia, 2020). The Information System Diploma Study Program has the following rules:

- The length of study at the diploma level in the Information System Study Program is between three to five years.
- Students who have exceeded the maximum study time are declared “drop out” (DO), with a decision letter that is issued by the head of the university at the suggestion of the dean of the faculty.

This is a consequence of educational regulations because of the large number of students. This is in line with research results that with limited education many students are often late in completing their studies and even drop out of school (Zuilkowski et al., 2019). An important problem faced by the Information System Diploma Study Program at Telkom University is the low timely graduation rate. Until 2016, the average diploma student who graduated on time was only 60%. This shows that the remaining 40% graduated late even dropped out. Delays in completing studies caused the burden of the learning process and the burden of lecturers to increase. In addition, the Study Program was also faced with the problem of difficulty in determining the final task that must be completed by students so that students found their own and completed the final project independently. To solve this problem, the department explored the problem that occurred. The Information System Diploma Study Program made observations to students regarding the problem of completing their final studies. The results of observations made by researchers on Class 2016 students showed that 59% had difficulty finding a final project topic for study completion, 32% did not know where to start in completing the final project, and 9% had other factors.

1.1 Conceptual Framework

Based on the problems above, we propose a new approach to the Information System Diploma Study Program, which is to group students' basic vocational skills by using data on certain subjects selected according to the 2010 Computing Curriculum (Isbell et al., 2009) and the SFIA Framework (J. Brown, 2020). Predictions made based on student course data at the beginning of the semester are then

mapped through the HWF (Hanung-Wawa-Ferra) Cycle on the dynamic program proposed by the Study Program. By doing the HWF cycle, the Information System Diploma Study Program has a program that can be a solution and help students complete their studies on time.

1.2 Related Works

There is quite a lot of researches related to grouping the abilities of students or people, such as research conducted by Antlová (2008). This research focuses on Information System competencies required by medium and small industries. The approach used was data mining methods, namely decision tree classifier, clustering, and association rules, to find competencies that are very influential on the long-term growth of a company. Another research is research conducted by Ariana et al. (2020). The purpose of this research is to measure the willingness of Information and Communications Technology (ICT) human resources to introduce e-government. This was done by introducing a resource-based view (RBV) to measure and classify ICT HR competence based on three indicators: technical competence, management competence and alignment. In addition, these indicators were measured using gap analysis and clustering techniques (Ariana et al., 2020).

Another interesting study is that of Lo, K.H et al. The research is about how the Hospital managers recognize the need to make their clinical staff more competitive, and patient-centered care models require innovation and continuous quality improvement to improve service quality and hospital competitiveness. The study uses two levels of grouping (Independent Organizational Map and K-Means) to identify different types of departments such as physicians, medical staff, and administrators, and to describe the core competencies of each group. This result shows how important it is to communicate core competencies to different hospital departments and units. (Kuei-Hsing Lo, Wen-Tsann Lin, Jou Yung-Tsan, 2018).

Other related research is research conducted by Ufuk Bolukbas (2017). The purpose of this study is to assess the technology management performance of SMEs so that comparisons can be made between firms in different clusters. K-Means cluster analysis was applied to the survey data to analyze and evaluate the performance of a manufacturing company in Istanbul, Turkey. As a result, a company's technological prowess is determined by 3, 4, and 5 clusters. Other related research is research conducted by Akarawang et al.(2015). This study aims to assess the needs and training to improve ICT competence. There were 377 teachers and 12 principals, supervisors and professional teachers from 35 schools were interviewed to identify issues and training needs. The results indicated that the teacher needed training that focused on acquiring practical skills and deepening her understanding of ICT.

Based on the related literature, there are several researchers who have the same thoughts in research, but research that specializes in grouping skills or competencies of Information System students as predictions to be mapped in a dynamic and sustainable study completion program has not been found. The HWF cycle as the basis of programming is something new as a contribution in this paper.

1.3 Purposes of the Study

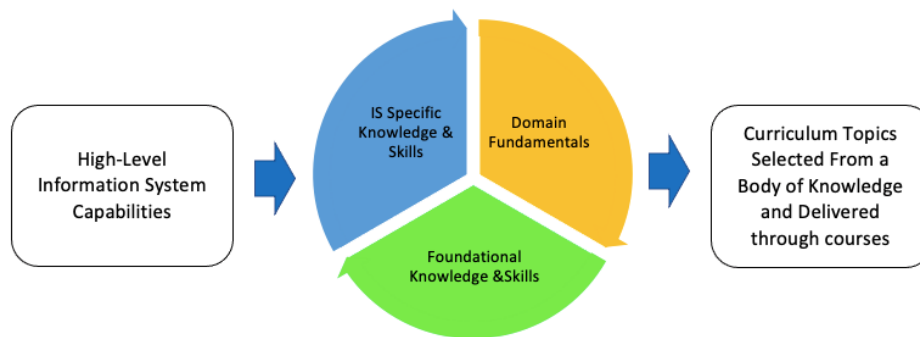
The purpose of this study is to produce a consistent model in measuring the competence of Diploma Program graduates on an ongoing basis in accordance with the characteristics of students in completing studies in the field of Information System.

2. Materials And Methods

2.1 Information System Vocational Competency

Vocational courses in the Information System Diploma Study Program at School of Applied Science of Telkom University are guided by the 2010 Computing Curriculum and the British SFIA framework (J. Brown, 2020; Isbell et al., 2009). This ability is based on knowledge and skills that have been categorized as specific knowledge and skills for Information System graduates. Figure 1 shows how skills are translated into curriculum documents.

Figure 1
High-Level Information System Capabilities



The use of the IS 2010 ACM Curriculum as the basis in this paper is because the data used is still based on the curriculum and further research is certainly needed for further development in the use of the IS 2020 curriculum (Isbell et al., 2009). Meanwhile, the SFIA Framework divides competencies into six categories as shown in Figure 2.

Figure 2
SFIA Framework (J. Brown, 2020)



Figure 2 shows that SFIA divides the competency areas into six main categories, namely: strategy and architecture, business change, solution development and implementation, service management, procurement and management support, and client interface.

2.2 Courses Selection

The selection of courses as a tool for predicting basic vocational skills and the selection of basic courses delivered in the first semester of student learning are adjusted to the curriculum that is run

based on the IS 2010 Computing Curriculum and the SFIA Framework. In addition, the concept of information system strategy is also considered, which includes Organization, Business, and Technology (Rahro & Soleymani, 2017). The overall mapping can be seen in Table 1.

Table 1
Mapping Framework used by the Department

No	IS 2010 Computing Curriculum	SFIA Framework	Basic Information System Competency Course
1	Improving organizational processes Seizing opportunities due to technological innovation Identifying and evaluating alternative solutions and sources	Strategy and Architecture, Service Management, Procurement and management support	Introduction to Management
2	Designing and managing enterprise architecture Understanding and addressing information needs	Business Change	Business Process
3	Protecting data and infrastructure, as well as Understanding, managing, and controlling risks.	Solution Development and implementation, Client interface	Information System Design and Analysis

2.3 K-Means Clustering

K-means clustering is a simple and elegant approach to partitioning a dataset into K distinct, non-overlapping clusters. To perform k-means clustering, we first need to determine the number of k-clusters we want. Then the k-means algorithm assigns exactly one cluster to each observation. (Bansal et al., 2017). The K-means algorithm is a non-hierarchical method that initially takes part of the population to be the initial cluster center. At this stage, the cluster center is randomly selected from a set of population data. Next, K-means tests each component in the data population and assigns that component to one of the predefined cluster centers depending on the minimum distance between the components and each cluster. The position of the cluster center will be recalculated until all data components are classified into each cluster center and finally a new cluster center position will be formed (Ahmad & Dey, 2007).

K-Means Clustering Algorithm provides easy implementation, very flexible, and easy to use adaptation. In addition to the time needed to do learning faster, it also uses simple principles that can be explained in non-statistical terms. This is the reason for using K-means clustering compared to other methods (Maylawati et al., 2020; Shabut et al., 2018). Determination of clusters in K-Means clustering using the Davies Bouldin Index (BDI) indicator. Davies Bouldin index (DBI) is a metric to evaluate or consider the results of clustering algorithms. By using DBI a cluster will be considered to have an optimal clustering scheme which has a minimum DBI (Karo et al., 2018).

3. Methodology

3.1 Research Model

This study used a quantitative approach. The flow of research carried out is as shown in the Figure 3

Figure 3

Research Flow



Figure 3 shows the flow of the research carried out. The steps taken were:

- **Problem Identification**
Problem identification is done to find and understand the main problems encountered. Identification is carried out thoroughly both for human problems (in this case, students and lecturers), identification in engineering problems, or related work structures and phenomena that occur (R Kumar, 2018).
- **Study Literature**
This stage is carried out for data collection techniques by conducting a review study of books, literatures, notes, and reports that have to do with the problem being solved (P. A. Brown, 2008).
- **Observation**
This stage needs to be done to strengthen the previous literature study stage. Observation also means observing and recording the subject or event in a systematic way. This is done to minimize errors in decision-making or modeling (Baker, 2006).
- **Program Modeling**
Modelling is the process of generating models. The model is a representation of the construction and work of some interesting systems. The model is made similar to but simpler than the system it represents. One of the goals of the model is to enable the analyst to predict the effects of changes on the system (Maria, 1997). The model is generated through observation and Forum Group Discuss with the Delphi method approach (Alarabiat & Ramos, 2019). This technique is very flexible to be applied in various situations and complex problems, in which there is often no suitable analytical method to be applied (Fink-Hafner et al., 2019).
- **Implementation**
At this stage, the implementation of the model is carried out to see the performance of the model as well as to evaluate the model.
- **Result**
The research results are obtained to evaluate the research that has been done and to find the potential for future development.

3.2 Participants

The research data used the data population of students of Information System Diploma Program at Telkom University in 2016.

3.3 Data Collecting Tool

The data was obtained from the academic system of Igracias at Telkom University. Igracias is academics Information system of Telkom University. The data used was data on the value of the subjects

that are the focus of the research, namely the Introduction to Management, Business Process, and IS Design & Analysis.

3.4 Data Collecting Process

The data collection process was carried out by filtering the data, selecting the year, and generating it on the Igracias Academic system at Telkom University.

3.5 Data Analysis

Analysis of the data in this study used the K-Means Clustering approach with the Davies Bouldin Index (BDI) indicator.

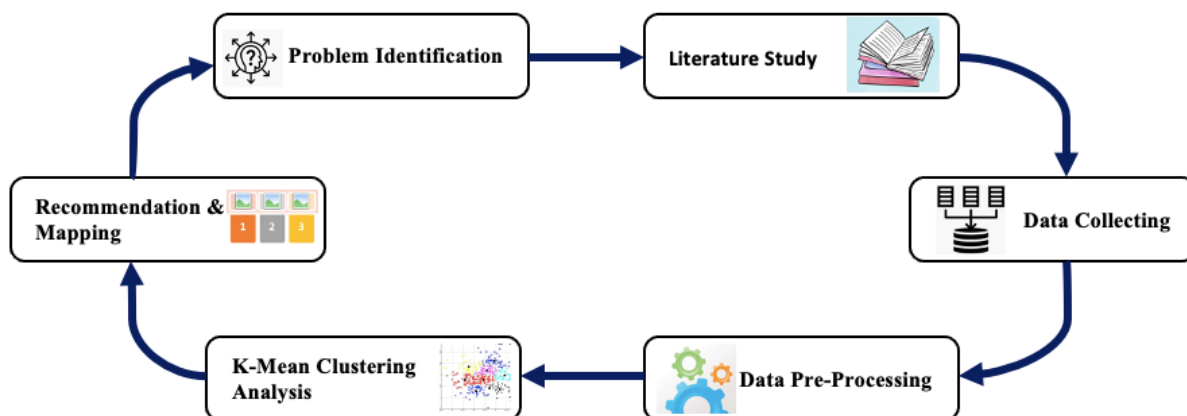
4. Result & Discussion

4.1 Program Modelling

The HWF Cycle model is generated based on the results of observations, literature study, and Forum Group Discuss (FGD) in the Information System Diploma Study Program, as shown in Figure 4.

Figure 4

HWF Cycle Model



The detailed flow of the HWF Cycle is as follows:

- Identifying the problem

This stage is to identify the problems faced by the Information System Diploma Study Program related to students' confusion and delays in completing studies.

- Literature review

At the literature review stage, a literature study is conducted related to research topics through information analysis of the basic competencies of Information System, analysis of the IS 2010 ACM Curriculum, scientific articles on the Scopus journal database, and so on.

- Data collection

The data collection stage is carried out after determining the elective courses that are the reference in the basic competencies of Information System. At this stage, data is filtered from the Igracias Academic information system database, Information System Diploma Study Program, School of Applied Sciences, Telkom University.

- Pre-processing

In this stage, pre-processing is carried out to make labels on research data.

- K-Means Clustering Analysis

At this stage, data processing and analysis are carried out using the K-Means Clustering approach.

- Recommendations

At this stage, mapping the selected clusters in the program proposed by Information System Diploma Study Program is dynamic. This is because each education force and student will have different conditions so that the problems faced will be different. Therefore, the results of the recommendations will always be evaluated every year and will change according to the problems encountered.

4.2 Data Processing Results

The initial data used previously is as shown in Table

Table 2

Secondary Data Obtained from the Igracias Academic System

NIM	Introduction to Management Score	Business Process Score	IS Design & Analysis Score
6701150001	64	68.25	60.25
6701150006	64	60	50.5
6701150025	70	41.25	65.7675
6701150032	73.75	61.6675	57.125
6701150034	79.75	71.75	80.25
...
6701164160	65	81.25	53.535
6701164162	79.4275	70	61.4375
6701164163	69.665	79.25	80.5

Table 3

Descriptive Statistics of initial data

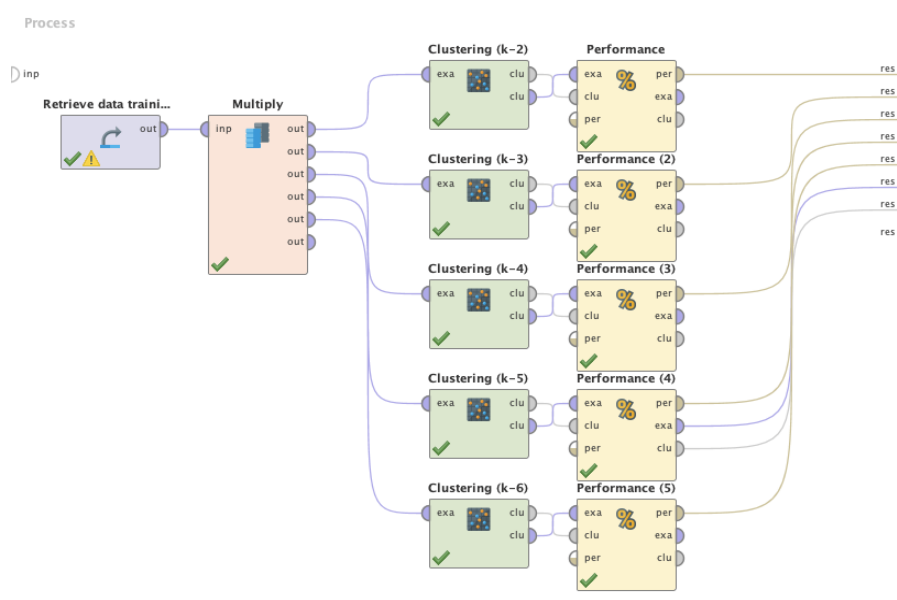
Course	N	Minimum	Maximum	Mean	Std. Deviation
Introduction to Management	251	35.43	90.00	72.8632	10.38931
Business Process	251	41.25	88.13	72.0032	6.92222
IS Design & Analysis	251	26.89	102.94	68.0514	11.88744
Valid N	251				

Data in Table 2 shows student competencies in the three core courses of the information System Diploma Study Program, namely Introduction to Management, Business Processes, and IS Design & Analysis. The three courses represent the level of student competence at each level of education. Student competencies at level 1 are represented by Introduction to Management, Business Processes, and IS Design & Analysis courses. Table 3 shows statistical data descriptively. The data shows that the course

that has the best average is Introduction to Management, followed by Business Process, and IS Design & Analysis.

By using Rapidminer, Figure 5 shows performance of the K-Means Clustering Model.

Figure 5
Performance of K-Means Clustering Model



In Figure 5, cluster selection is carried out by performing five iterations. The first iteration (Clustering (k-2)) is selecting two clusters and then obtaining the Davies Bouldin value. The second iteration (Clustering (k-3)) is selecting three clusters until the fifth iteration (Clustering (k-6)) with the number of selected clusters is 6. Determination of the number of clusters is based on Davies Bouldin values, which is based on data processing using Rapidminer. Davies Bouldin values obtained is as shown in Table 4.

Table 4
Davies Bouldin Values in K-Means Clustering Iteration

No	Number of K	Davies Bouldin's Value
1	2	1.148
2	3	1.074
3	4	1.120
4	5	1.054
5	6	1.061

Based on the Davies Bouldin values, the number of selected clusters is 5. The centroid value of the 5 selected clusters can be seen in Table 5.

Table 5
Centroid Value of Clusters

Attribute	Cluster_0	Cluster_1	Cluster_2	Cluster_3	Cluster_4
IS Design & Analysis Score	76.735	65.572	63.705	79.092	47.518
Introduction to Management Score	82.011	49.899	75.772	67.567	65.905

Business Process Score	76.825	75.785	68.862	73.032	66.641
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The results of the cluster grouping are as shown in Table 6.

Table 6

Distribution of Data on the 5 Selected Clusters

Cluster	Items
0	69
1	18
2	86
3	45
4	33
Totally	251

The cluster grouping is as shown in Table 7-11

Table 7

Cluster_0

No	NIM	Introduction to Management Score	Business Process Score	IS Design & Analysis Score
13	6701150034	79.75	71.75	80.25
18	6701150041	82	87.5	72.25
29	6701150108	87.5	71.55	71.25
30	6701150109	83.75	77.35	91.75
...
238	6701164123	83.0325	76	69.25
248	6701164155	90	77.25	67.3225

Table 8

Cluster_1

No	NIM	Introduction to Management Score	Business Process Score	IS Design & Analysis Score
136	6701160011	51.3225	80	73.75
141	6701160023	54.68	76	70.75
179	6701160114	41.635	71.25	65
180	6701160115	61.5	73.75	61.1425
...
234	6701164079	45.615	77.5	67
245	6701164141	51.645	73.75	64.875

Table 9

Cluster_2

No	NIM	Introduction to Management Score	Business Process Score	IS Design & Analysis Score
1	6701150001	64	68.25	60.25

4	6701150011	72	73.75	55.25
5	6701150015	80	58.125	60.92
10	6701150025	70	41.25	65.7675
...
246	6701164147	65.885	70	63.75
250	6701164162	79.4275	70	61.4375

Table 10
Cluster_3

No	NIM	Introduction to Management Score	Business Process Score	IS Design & Analysis Score
7	6701150019	67.5	64.6	77
27	6701150106	68	73.75	71.75
60	6701151062	68.75	64.1675	77.75
93	6701154094	73.75	69.25	79.5
...
247	6701164149	68.2675	65	75.625
251	6701164163	69.665	79.25	80.5

Table 11
Cluster_4

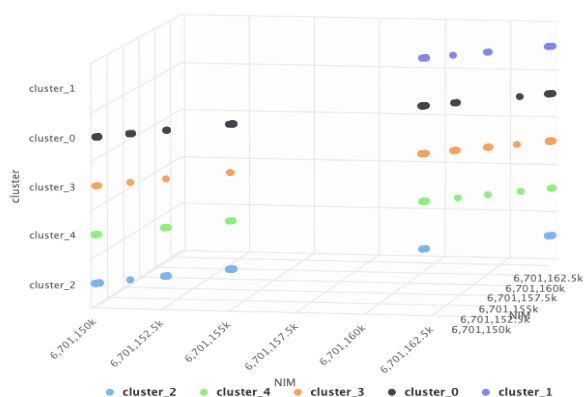
No	NIM	Introduction to Management Score	Business Process Score	IS Design & Analysis Score
2	6701150006	64	60	50.5
3	6701150010	65	50.9375	52.1425
15	6701150038	83.75	78.35	41
...
244	6701164140	53.25	66.25	43.25
249	6701164160	65	81.25	53.535

The main purpose of the clustering process is to group objects or cases based on their characteristics. Cluster analysis classifies objects so that each object that has similar properties (closest in similarity) will be grouped into the same cluster (group). In this study, clusters were conducted to see groups of students who have the same characteristics in terms of basic information system abilities.

4.3 Clustering Analysis

The results of data processing using Rapidminer showed that the best cluster was cluster 1, as shown in Figure 6.

Figure 6
Cluster Diagram of Data Processing Results



Based on the performance of K-Means Clustering as shown in Figure 6 above, it can be seen that the order of clusters from the best is cluster_1, cluster_0, cluster_3, cluster_4, and cluster_2.

4.4 Discussions

Five clusters were obtained in this study. This grouping can be a recommendation for the Information System Diploma Study Program to create a completion program for Information System Diploma Program. The 5 programs compiled are based on the abilities possessed by 5 student clusters, which are based on the basic competencies of Information System Diploma Program.

Table **12**
Mapping of Proposals Based on the Results of Clustering.

No	Cluster	Program Level	Description
1	Cluster_1	Superior	Students understand management concepts, are able to reengineer business processes, build information technology-based information system, and complete large industrial application development programs. This level is the highest level because the level faced is relatively difficult.
2	Cluster_0	Primary	Students understand management concepts, are able to reengineer business processes, build information technology-based information system, and complete small industry application development programs. At this level, the level of difficulty is high and students work in groups in the information system development team.
3	Cluster_3	Secondary	Students understand management concepts, are able to reengineer business processes, build information technology-based information system, and complete government application development programs. This level is included in the middle level category because the applications built are based on processes that generally have a similar model compared to the two previous levels.
4	Cluster_4	Tertiary	Students understand management concepts, are able to reengineer business processes, build information technology-based information system, and complete lecturer

No	Cluster	Program Level	Description
5	Cluster_2	Non-Superior	<p>research application development program. At this level, the level of difficulty is low because the research carried out can be adjusted to student's ability in accordance with the subjectivity of the lecturer as a researcher.</p> <p>Students understand management concepts, are able to reengineer business processes, build information technology-based information system, and complete the simple application development program. This level is the lowest level that can be completed by students according to their abilities and is free but still adjusts to the minimum standards of the Study Program.</p>

Through the adjustment of the number of programs that is dynamic, Table 12 shows the adjustment of the program with the results of the cluster of information system basic competencies. The indicator of program difficulty level is based on the results of FGDs from expert judgment lecturers, industry partners, and the government that collaborate with the Information System Diploma Study Program.

4.5 Recommendations

After clustering students based on their competencies, a suitable treatment program is recommended for each student cluster. Table 13 is a recommendation for each cluster of students in order to graduate on time.

Table 13
Mapping of proposals based on the results of clustering.

No	Program Level	Competency Improvement Program	Motivation Improvement Program
1	Superior	Students are trained to have the ability in time management and project management so that they can optimize their capabilities and resources in a directed and effective manner to achieve the target on time before the deadline for graduating from study.	<ul style="list-style-type: none"> • Emotional control program • Creative thinking • Commitment to time • Work effectively and efficiently • Communication and collaboration
2	Primary		
3	Secondary		
4	Tertiary	Students are trained to be more skilled in the development of software and information system and have the ability in time management and project management so that they can compete with other students to complete larger projects.	<ul style="list-style-type: none"> • Emotional control program • Creative thinking • Commitment to time • Work effectively and efficiently • Communication and collaboration • Passion for self-development • Not easily discouraged • Positive thinking
5	Non-Superior		

4.6 Evaluation

The program that had been implemented was significantly quite successful. This can be seen from the increase in the timely graduation rate of Information System Diploma Study Program as shown in Figure 7.

Figure 7

Graduation Rate on Time of Students

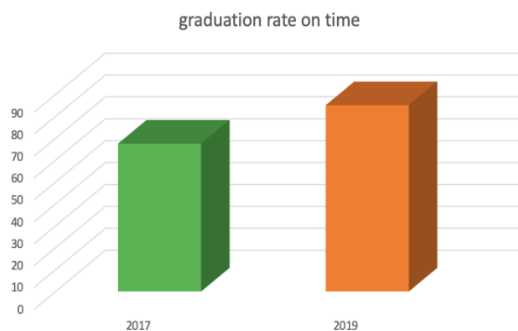


Figure 7 shows that before running this program, the timely graduation rate in the Information System Diploma Study Program only reached 67.48%. Meanwhile, after running this program, the timely graduation rate experienced a very significant increase of 85.04%.

5. Conclusion & Future Works

Based on the results of the study, it was concluded that the K-Means clustering mechanism in the HWF Cycle facilitated the Study Program to develop a final project completion program so that students were more focused in completing their final studies. The dynamic nature of the program will be an evaluation material for the Information System Diploma Study Program in determining the program that suits the needs of students in completing their final studies, which is based on predictions of initial data on the students' basic competencies of Information System Diploma Program. The program that had been run showed significant results. Before the program was run, the timely graduation rate was only 67.48% in 2017. Meanwhile, after the program was run, the timely graduation rate reached 85.04% in 2019. This means that there was an increase of 26%.

This research was conducted locally and only applies to Information System Diploma Study Program. For further research, it can be improved through larger student data from various universities that have the same major and use the latest Information System Curriculum approach, namely IS 2020 and The New SFIA Framework.

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Declaration of Conflicting Interests

The authors declared no potential conflicts of interest with respect to the research, authorship, and/or publication of this article.

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