

1. Paper submitted to TELKOMNIKA (23-07-2019)
2. Paper accepted with minor revision (22-12-2019)
3. Paper revised (21-01-2020)
4. Paper Proofreading [TELKOMNIKA Proofreading] Paper ID 14802 (15-03-2020)
5. Paper published (1-06-2020)

1. Paper submitted to TELKOMNIKA (23-07-2019)



---

## [TELKOMNIKA] Submission Acknowledgement "Predicting student performance using data mining"

---

Tole Sutikno <telkomnika@uad.ac.id>  
To: Dr Leo Willyanto Santoso <leow@petra.ac.id>

Fri, Jul 23, 2019 at 11:18 AM

Dear Prof/Dr/Mr/Mrs: Dr Leo Willyanto Santoso:

Thank you for submitting the manuscript, "Predicting student performance using data mining" to TELKOMNIKA (Telecommunication Computing Electronics and Control). With the online journal management system that we are using, you will be able to track its progress through the editorial process by logging in to the journal web site:

Manuscript URL:

<http://journal.uad.ac.id/index.php/TELKOMNIKA/author/submission/14802>

Username: leow

Your article ID is number from the URL:

<http://journal.uad.ac.id/index.php/TELKOMNIKA/author/submission/14802>

If you have any questions, please contact me. Thank you for considering this journal as a venue for your work.

Best Regards,

Tole Sutikno

TELKOMNIKA (Telecommunication Computing Electronics and Control)

-----  
The best journal in Indonesia 2017-2019 (by Ministry of Research Technology and Higher Education) and indexed by SCOPUS, with CiteScore 2016: 0.75, SJR 2016: 0.250 & SNIP 2016: 0.822

-----  
----- Our events -----  
-----

An International Conference Indexed by IEEEExplore and Scopus

"4th IEEE Conference on Energy Conversion (CENCON 2019)"

Website: <http://cencon.intconference.org>

EDAS papers submission: <https://edas.info/N25880>

The 4th IEEE Conference on Energy Conversion (CENCON 2019) will be held at Royal Ambarrukmo hotel, located in the capital of Yogyakarta, Indonesia on 16-17 October 2019. The objective of the conference is to share the latest research in the areas of power electronics, electrical drives, renewable energy, modeling and control systems. CENCON 2019 is sponsored by the IEEE Malaysia Power Electronics (PEL) Chapter, co-organized by IEEE Indonesia Power Electronics (PEL) Chapter and Universitas Ahmad Dahlan and technical co-sponsored by The Korean Institute of Power Electronics (KIPE) and Power Electronics and Drives Research Group (PEDG).

---

2019 6th International Conference on Electrical Engineering, Computer Science and Informatics (EECSI 2019), September 18-20, 2019 in Bandung, Indonesia.

<http://eecsi.org/2019>

The conference is hosted by Universitas Budi Luhur and is jointly organized with Universitas Gadjah Mada, Universitas Diponegoro, Universitas Sriwijaya, Universitas Ahmad Dahlan, Universitas Islam Sultan Agung, Universitas Muhammadiyah Malang, Universiti Teknologi Malaysia, and IAES Indonesia Section. The event is intended to provide technical forum and research discussion related to advanced engineering on electrical & electronics, computer science and informatics.

-----  

---

**TELKOMNIKA (Telecommunication Computing Electronics and Control)**

<http://www.journal.uad.ac.id/index.php/TELKOMNIKA>

2. Paper accepted with minor revision (22-12-2019)



---

## [TELKOMNIKA] Editor Decision "Predicting student performance using data mining"

---

Tole Sutikno <telkomnika@uad.ac.id>  
Reply-To: "Dr. Tole Sutikno" <telkomnika@uad.ac.id>  
To: Dr Leo Willyanto Santoso <leow@petra.ac.id>

Sun, Dec 22, 2019 at 9:51 PM

Dear Prof/Dr/Mr/Mrs: Dr Leo Willyanto Santoso,

It is my great pleasure to inform you that your paper entitled "Predicting student performance using data mining" has been accepted and will be published on the TELKOMNIKA Telecommunication Computing Electronics and Control (ISSN 1693-6930, SCOPUS/ScimagoJR indexed journal, Q2 on Electrical and Electronics Engineering, SJR: 0.283, CiteScore: 1.09, SNIP: 0.730). Congratulations!

Please prepare your final paper (in MS Word file format) adheres every detail of the guide of authors, and check it for spelling/grammatical mistakes.

In order to cover part of the publication cost, each accepted paper will be charged:

- \* USD 265 (for Overseas Authors)
- \* IDR 3500K (for Indonesian Authors)

This charge is for the first 8 pages, and if any published manuscript over 8 pages will incur extra charges USD 50 (or IDR 600K for Indonesian Authors) per page

The payment should be made by bank transfer (T/T):

-----  
Bank Account name (please be exact)/Beneficiary: TOLE SUTIKNO  
Bank Name: Bank Central Asia (BCA)  
Branch Office: Kusumanegara  
City: Yogyakarta  
Country :Indonesia  
Bank Account #: 8465122249  
SWIFT Code: CENAIJAXXX

or through PayPal (as alternative of bank transfer) to email:  
[tole@ee.uad.ac.id](mailto:tole@ee.uad.ac.id)

-----  
Your paper will be scheduled for forthcoming issue. Please pay the publication fee as soon as possible (within 3 weeks). If you need more time, please send a request to this email. We can give you 5 weeks at the most.

Then, please submit your final paper & payment receipt to this email

I look forward for your response

Sincerely yours,  
Dr. Tole Sutikno

Universitas Ahmad Dahlan  
[telkomnika@uad.ac.id](mailto:telkomnika@uad.ac.id)

-----  
Bank's detailed address :  
Bank BCA Kusumanegara  
Jl. Kusumanegara No. 18  
City: Yogyakarta  
Province: D.I. Yogyakarta (DIY)  
Country :Indonesia  
Post Code: 55165  
Indonesia, Phone:+62 274 418896

The Beneficiary's address:  
Kampus 3 Universitas Ahmad Dahlan  
Jln. Prof. Soepomo, Janturan  
City: Yogyakarta  
Province: D.I. Yogyakarta (DIY)  
Post Code: 55164  
Country: Indonesia

=====

TELKOMNIKA Telecommunication, Computing, Electronics and Control is a Scopus indexed leading journal, Scimago journal ranking (SJR) Q2 on Electrical & Electronics Engineering. This journal ONLY publishes high quality papers. A high quality paper MUST has: 1) a clear statement of the problem the paper is addressing; 2) the proposed solution(s); and 3) results achieved. It describes clearly what has been done before on the problem, and what is NEW. The goal of your final camera ready paper is to describe NOVEL TECHNICAL RESULTS.

Update title of your paper!! The title summarizes the main idea or ideas of your study (title is "summary" and "essence" of your paper). Title should be "encompassing" as well as "descriptive". A good title contains the fewest possible words needed to adequately describe the content and/or purpose of your research paper. Rarely use abbreviations or acronyms unless they are commonly known. Find the below guide, how to update your paper title.

For original research paper, there are four (4) types of novel technical results: 1) An algorithm; 2) A system construct: such as hardware design, software system, protocol, etc.; The main goal of your revised paper is to ensure that the next person who designs a system like yours doesn't make the same mistakes and takes advantage of some of your best solutions. So make sure that the hard problems (and their solutions) are discussed and the non-obvious mistakes (and how to avoid them) are discussed; 3) A performance evaluation: obtained through analyses, simulation or measurements; or 4) A theory: consisting of a collection of theorems. Your final camera ready paper should focus on: 1) Describing the results in sufficient details to establish their validity; 2) Identifying the novel aspects of the results, i.e., what new knowledge is reported and what makes it non-obvious; and 3) Identifying the significance of the results: what improvements and impact do they suggest. Number of minimum references for original research paper is 25 references (and minimum 20 recently journal articles).

For review paper, the paper should present a critical, constructive analysis of the literature in a specific field through summary, classification, analysis and comparison. The function and goal of the review paper is: 1) to organize literature; 2) to evaluate literature; 3) to identify patterns and trends in the literature; 4) to synthesize literature; or 5) to identify research gaps and recommend new research areas. The structure includes:

1. Title – in this case does not indicate that it is a review article.
2. Abstract – includes a description of subjects covered.
3. Introduction includes a description of context (paragraph 1 – 3), motivation for review (paragraph 4, sentence 1) and defines the focus (paragraph 4, sentences 2 – 3)
4. Body – structured by headings and subheadings
5. Conclusion – states the implications of the findings and an identifies possible new research fields
6. References (“Literature Review”) – organised by number in the order they were cited in the text.

Number of minimum references for review paper is 50 references (and minimum 40 recently journal articles).

Prepare your abstract in single paragraph and within 200 words. You need to summarize your contribution, idea, findings/results, and describe implications of the findings. Without abbreviations, footnotes, or references. Without mathematical equations, diagram or tabular material. It is suggested to present your abstract included the elements: 1) state the primary objective of the paper; 2) highlight the merits (or contribution; 3) give a conceptual idea on the method; 4) describe the research design and procedures/processes employed (is it simulation, experimental, survey etc.); 5) give the main outcomes or results, and the conclusions that might be drawn; and 6) include any implications for further research or application/practice, if any.

Please refer to <https://bit.ly/35R6JTs> and <https://bit.ly/2DxU9MI> for further guidelines

Your final camera ready paper should reflect a careful consideration of the following criteria:

1. Analysis: your revised paper should demonstrate a clear understanding of the key issues related to your topic of choice. The paper should display analysis and not mere summary of the topic under consideration. It should also include evidence to support arguments where necessary.
2. Connections: your paper should demonstrate a connection of the references you mention to the central topic and to each other where necessary throughout the paper.
3. Mechanics: this includes attention to punctuation, grammatical soundness and your submissions being checked for spellings errors.
4. Formatting: adhere the new guide of authors (<http://icw.telkomnika.com/download/telkomnika-template-and-guide-of-authors-2020/>)

-----  
Guideline for preparing your paper title:  
-----

A good research paper title: (1) Condenses the paper’s content in a few words & Use words that create a positive impression and stimulate reader interest; (2) Captures the readers’ attention; (3) Indicate accurately the subject and scope of the study and Differentiates the paper from other papers of the same subject area. Five (5) Simple steps to write a good research paper title:

STEP 1: Ask yourself these questions and make note of the answers:

- What is my paper about?



- What techniques/ designs were used?
- Who/what is studied?
- What were the results?

STEP 2: Use your answers to list key words.

STEP 3: Create a sentence that includes the key words you listed

STEP 4: Delete all unnecessary/ repetitive words and link the remaining.

STEP 5: Delete non-essential information and reword the title.

-----  
 General Guidelines:  
 -----

1. Please re-read our instructions (at:

<http://journal.uad.ac.id/index.php/TELKOMNIKA/about/editorialPolicies#custom-1>)

carefully and follow the checklist strictly, as any spelling mistakes and errors may be translated into the typeset version.

2. The “result and discussion” section reports the most important findings, including analysing results as appropriate. It is very important to prove that your manuscript has a significant value and not trivial.

3. Please re-check that all references are already cited in your article, and order of your citation is SEQUENTIAL

example in a paper:

—> [1-4], [2], [5-6], [7-9], [8], [4-5], [9], [10-14], [12], [15] .....

(SEQUENTIAL) — correct

—> [1], [2], [3], [4-6], [7], [8-10], [4-5], [11-16], [13], [17] .....

(SEQUENTIAL) — correct

—> [2], [3], [6], [1], [4-5], [7], [11-16], [8-10], [13], [17] .....

(NOT SEQUENTIAL) — INCORRECT

—> [4-5], [7], [2], [3], [17], [6], [1], [11-16], [8-10], [13] .....

(NOT SEQUENTIAL) — INCORRECT

4. If you need references to enrich your references and improve your paper, please get take a look at:

- <http://beei.org>
- <http://ijece.iaescore.com>
- <http://ijeecs.iaescore.com>
- <http://journal.uad.ac.id/index.php/telkomnika>
- <http://ijpeds.iaescore.com>
- <http://ijai.iaescore.com>

Please use “Search” at menu "Journal Content" in right side of the site.

---

TELKOMNIKA (Telecommunication Computing Electronics and Control)

<http://www.journal.uad.ac.id/index.php/TELKOMNIKA>

3. Paper revised (21-01-2020)



UNIVERSITAS  
KRISTEN  
PETRA

Leo Willyanto <leow@petra.ac.id>

---

## [TELKOMNIKA] Editor Decision "Predicting student performance using data mining"

---

Leo Willyanto <leow@petra.ac.id>  
To: "Dr. Tole Sutikno" <telkomnika@uad.ac.id>

Tue, Jan 21, 2020 at 10:41 AM

Dear Prof. Dr. Tole,

Sorry for the late reply.  
Attached are my paper and payment receipt. Thank you.

Leo W. Santoso  
[Quoted text hidden]

### 2 attachments

**Bayar Jurnal.jpeg**  
66K

 **Leo-Telkomnika Rev.docx**  
670K



UNIVERSITAS  
KRISTEN  
PETRA

Leo Willyanto <leow@petra.ac.id>

---

## [TELKOMNIKA] Editor Decision "Predicting student performance using data mining"

---

Jurnal TELKOMNIKA <telkomnika@uad.ac.id>  
To: Leo Willyanto <leow@petra.ac.id>

Tue, Jan 28, 2020 at 11:11 PM

Your paper ID is **14802**

Best Regards,  
Tole Sutikno  
*Editor-in-Chief, TELKOMNIKA (Telecommunication, Computing, Electronics and Control)*  
ISSN: 1693-6930, e-ISSN: 2302-9293  
<http://journal.uad.ac.id/index.php/TELKOMNIKA>

[Quoted text hidden]  
[Quoted text hidden]

[Quoted text hidden]

4. Paper Proofreading [TELKOMNIKA Proofreading] Paper ID 14802 (15-03-2020)



---

## [TELKOMNIKA Proofreading] June 2020 Paper ID 14802

---

Hany Safitry F <hanyjournal.iaes@gmail.com>  
To: Leo Willyanto <leow@petra.ac.id>

Sun, Mar 15, 2020 at 5:54 PM

Dear Sir/Madam,

We hope this email reaches you well. We are from TELKOMNIKA Staff want to inform you that your paper **ID 14802 entitled Predicting student performance in higher education using multi-regression models will be published in June 2020 Issue**. We suggest you to checking your paper that attaches in this email.

If you have something to tell us about your paper, please reply to this email and comment on your pdf. We wait for your responses. We will wait for your responses until March 17, 2020. If you have passed the deadline, then we consider it to be the final paper. Thank you for your cooperation.

Best Regards,  
Hany Safitry.  
TELKOMNIKA Staff

### TELKOMNIKA Journal

Telecommunication, Computing,  
Electronics, and Control



Editor in Chief : [telkomnika@uad.ac.id](mailto:telkomnika@uad.ac.id)  
Staff : [hanyjournal.iaes@gmail.com](mailto:hanyjournal.iaes@gmail.com),  
cc: [enengnura3n1@gmail.com](mailto:enengnura3n1@gmail.com)

Universitas Ahmad Dahlan, 4th Campus, 9th Floor, LPPI Room  
Ringroad Selatan St., Kragilan, Tamanan, Banguntapan, Bantul,  
Yogyakarta, Indonesia 55191

Website :  
[journal.uad.ac.id/index.php/TELKOMNIKA/index](http://journal.uad.ac.id/index.php/TELKOMNIKA/index)

---

### OUR EVENTS

---

#### **2020 1st Conference on Internet of Things and Embedded Intelligence (CITEI 2020)**

Venue: Yogyakarta, Indonesia

Date: July 29-30, 2020

Website: <http://citei.intconference.org>

Link of Paper submission: <https://edas.info/N27031>

Paper submission Deadline: February 20, 2020

---

**2020 2nd ADICS Int Conference on Engineering, Science and Information Technology**

Venue: Yogyakarta, Indonesia

Date: August 26-27, 2020

Website: <http://esit.intconference.org>

Link of Paper submission: <https://edas.info/N27091>

Paper submission Deadline: March 20, 2020

---

**2020 7th International Conference on Electrical Engineering, Computer Science and Informatics (EECSI 2020)**

Venue: Yogyakarta, Indonesia

Date: September 23-25, 2020

Website: <http://eecsi.org/2020/>

Link of Paper submission: <https://edas.info/N27101>

Paper submission Deadline: April 15, 2020

---

**2020 3rd International Conference and Workshop on Telecommunication, Computing, Electrical, Electronics and Control (ICW-TELKOMNIKA 2020)**

Venue: Yogyakarta, Indonesia

Date: November 19-21, 2020, 2020

Website: <http://icw.telkomnika.com>

Link of Paper submission: <http://journal.uad.ac.id/index.php/TELKOMNIKA> (section ICW....)

Paper submission Deadline: June 15, 2020

---



27 14802 (23Jul19) (21Jan20) (24Feb20) atika.pdf

637K



UNIVERSITAS  
KRISTEN  
PETRA

Leo Willyanto <leow@petra.ac.id>

---

**[TELKOMNIKA Proofreading] June 2020 Paper ID 14802**

---

**Leo Willyanto** <leow@petra.ac.id>  
To: Hany Safitry F <hanyjournal.iaes@gmail.com>

Tue, Mar 17, 2020 at 12:30 PM

Dear Hany Safitry,

Yes, please proceed my article. Thank you.  
[Quoted text hidden]



UNIVERSITAS  
KRISTEN  
PETRA

Leo Willyanto <leow@petra.ac.id>

---

**[TELKOMNIKA Proofreading] June 2020 Paper ID 14802**

---

**Hany Safitry F** <hanyjournal.iaes@gmail.com>  
To: Leo Willyanto <leow@petra.ac.id>

Tue, Mar 17, 2020 at 2:05 PM

Dear Sir/Madam,

Thank you for your confirmation. We will upload to our system as soon as possible.  
[Quoted text hidden]  
[Quoted text hidden]

[Quoted text hidden]  
[Quoted text hidden]

## Predicting student performance in higher education using multi-regression models

Leo Willyanto Santoso, Yulia

Informatics Department, Petra Christian University Surabaya, Indonesia

### Article Info

#### Article history:

Received Jul 23, 2019

Revised Jan 21, 2020

Accepted Feb 24, 2020

#### Keywords:

Data mining

Education

Multi-regression

Prediction

Student

### ABSTRACT

Supporting the goal of higher education to produce graduation who will be a professional leader is a crucial. Most of universities implement intelligent information system (IIS) to support in achieving their vision and mission. One of the features of IIS is student performance prediction. By implementing data mining model in IIS, this feature could precisely predict the student' grade for their enrolled subjects. Moreover, it can recognize at-risk students and allow top educational management to take educative interventions in order to succeed academically. In this research, multi-regression model was proposed to build model for every student. In our model, learning management system (LMS) activity logs were computed. Based on the testing result on big students datasets, courses, and activities indicates that these models could improve the accuracy of prediction model by over 15%.

*This is an open access article under the [CC BY-SA](#) license.*



### Corresponding Author:

Leo Willyanto Santoso,

Informatics Department,

Petra Christian University Surabaya,

121-131 Siwalankerto St., Surabaya, East Java 60236, Indonesia.

Email: leow@petra.ac.id

## 1. INTRODUCTION

Education is a key to ending the poverty in developing countries. Education has power to change the people, communities, nation and human life. The government should pay more attention to the quality of education. Education is the responsibility of the stakeholders including government official, parent, and teacher. Education should be managed through national resources. Furthermore, higher education is important for social and economic impacts in society. The general mission of higher education institution is to produce student graduation who will be a professional leaders in their field and valuable for their communities and country. To achieve this mission, higher education institution should improve their quality of education. There are several factors affected the quality of education. The high level of student success and low failure rate students can reflect the quality of education. One of the major problems of higher education in the developing country, like Indonesia is the high rates of student drop out that has reached 10%. Another related problem is the long time that a student takes to complete their degree. Nowadays, information technology is considered as important factor to improve the quality of education. This is the reason why many universities are investing a lot of budget to improve their academic information system [1].

Educational data mining (EDM) has emerged in the last decades due to the large volume of educational data that was made available [2, 3]. It is concerned with developing and applying data mining algorithms to identify patterns in large amounts of educational data, and to better understand students and their learning environments [4-7]. Moreover, data mining and data warehousing technique have been increasingly implemented in the academic information system to analyze the vast amounts of student data [8, 9].



Data mining is a tool to improve the quality of education by identifying the students who are at risk in their study [10-12]. This information is very useful for top level management to take appropriate action for students who are considered to have a higher probability of failing academically or dropping out of university. The university could provide additional services and resources to the at-risk students [13, 14]. In addition, they need to develop innovative approaches to retain students, ensure that they graduate on a timely manner [15].

Some techniques have been developed to address this issue. However, these approaches ignore the different features of how students work together with the material/LMS' provided information, which could possibly be used to increase overall accurateness of prediction. In this research, single regression model and multi regression model were implemented and investigated. This model could predict the students' grade by mining different course activities log (e.g., tests and assignments) in learning management system. An early warning system generates early warnings about struggling students who are most likely to failed a course or drop out of university. It is supposed to generate these warnings early enough in order to allow for intervention by offering suitable assistance for the students that are at risk. This system works by predicting a student's performance in the learning activities (e.g., assignments) within a course that they are enrolled in. They also predict the student's final grade in a course that they are enrolled in, or in courses that they will take in the next semester to fulfill their program requirements.

When students first enroll in a university, their university get the data about their performance in various high school subjects, test academic potential, and demographics. As the students proceed with their academic studies, more data are collected. The collected data like the student transcript and enrolled courses. The students can also access online learning management system (LMS), such as Moodle, Edmodo, Eliademi, ATutor or BlackBoard, at which they get access to the course materials. Through the LMS, students can also engage in forum discussions, contribute to the course content, engage in course activities such as online quizzes, and do other tasks. In this research, large dataset was extracted from the Petra Christian University's LMS. The name of Petra Christian University LMS is Lentera, based on Moodle [16]. This dataset contains 486 courses, 7,563 students, and 109,231 activities.

The important contributions of this paper are as follows: (1) The designed system can cluster/segment the students into groups whose prediction models are relatively similar. By exploring these student' groups, knowledge on the factors that determine the students' performance are gained. (2) The proposed recommender system provides solution to improve the education quality using cutting edge technology. The rest of the paper is organized as follows: section 2 describes the literature review. Section 2 describes the multi-regression model that we used. Section 2 describes the dataset that we used along with the various features that we extracted. Section 3 provides the investigational evaluation and analysis of the results. Finally, section 4 concludes this research.

## 2. RESEARCH METHOD

Identifying at-risk students for taking appropriate actions can be addressed through evaluating collected students' academic performance data. Decision tree technique was implemented to explain the properties interdependencies of drop out students [17]. This study also offers an example of how data mining technique can be used to increase the effectiveness and efficiency of the modeling processes. Dekker explained a data-mining case study demonstrating the usefulness of several classification methods and the cost-sensitive learning approach [18]. In this system, cost-sensitive learning does help to bias classification errors towards preferring false positives to false negatives. Optimization should be done to improve the system.

Predictive analytic technique could be integrated with learning management system (LMS) to identify students who are in danger of failing the course in which they are currently enrolled [19]. Learning analytic is considered can support students, lecturers and educational managers to predict course failure [20]. Learning analytic be able to support instructional material designers to better measure the quality of a course design and understand what works and what does not work [21, 22]. Moreover, learning analytic can increase evaluation of student performance by investigating various indicators such as student activities and grades on assignments.

Data mining techniques for categorizing university students based on Moodle' usage data in a learning management system and the final marks achieved in the course was implemented [23]. The proposed system uses preprocessing tasks as discretization and rebalancing data. The author should consider how the data quantity and data quality can impact the performance of the algorithms. Information with more evidence about the students, like student profile and set of courses should be incorporated.

Tensor factorization techniques for predicting student performance was proposed [24]. The author introduces a novel recommender system which can be used not only for recommending objects like tasks/exercises to the students but also for predicting student performance. The prediction results could be improved by applying more sophisticated methods to deal with the cold-start problems and building ensemble methods on different models generated from matrix and tensor factorization.

Several factors effecting the accomplishment of the freshman students was determined [25]. The developed system can classify students into three groups: ‘low-risk’ students, with a high probability of succeeding; ‘medium-risk’ students, who may succeed; and ‘high-risk’ students, who have a high probability of dropping out. However, the combination of different prediction methods have not been addressed. This combination may lead to the improvement of the overall result.

With large volumes of student data, including enrollment, academic and disciplinary records, higher education institution could build big data and analytics system [26]. Big data can provide top level management the needed analytical tools to improve learning output for individual students as well ways guaranteeing academic programmes are of high-quality standards [27]. By designing applications that gather data at every phase of the students learning processes, universities can address student needs with customized modules, feedback, and assignments in the syllabus that will stimulate better and richer learning. In this research, we investigate the linear multi-regression models to forecast the students’ performance at various course activities in LMS.

### 2.1. Design

In this part, the proposed model for prediction student performance will be discussed. This model uses multi-regression model [28, 29]. Multi-regression is an extension of simple linear regression. As a predictive analysis, the multi-regression is used to explain the relationship between dependent variable and two or more independent variables. In this model, the grade  $g_{sa}$  for student  $s$  in activities  $a$  is formulated as (1).

$$\begin{aligned} g_{sa} &= b_s + b_c + \sum_{s=1}^l p_s W f_{sa} \\ &= b_s + b_c + \sum_{d=1}^l (p_{s,d} \sum_{k=1}^{n_f} f_{sa,k} w_{d,k}) \end{aligned} \quad (1)$$

where:

$b_s$  = student bias terms

$b_c$  = course bias terms

$f_{sa}$  = vector that stores the input features

$l$  = total of linear regression models

$W$  = matrix that stores the coefficients of linear regression

$p_s$  = vector that stores the memberships of student  $s$

$w_{d,k}$  = weighted feature  $k$  under the  $d^{\text{th}}$  regression model

$p_{s,d}$  = student membership  $s$  in the  $d^{\text{th}}$  regression model

The performance comparison between a multi-regression model across a linear regression model was presented. The approximation of university student grade using linear regression model as;

$$g_{sa} = w_0 + \sum_{k=1}^{n_f} w_k f_k \quad (2)$$

where  $f_k$  is the rate of  $k$  and the  $w_k$ 's are the regression coefficients.

In Figure 1 can be seen the flow diagram of application design process. The initial stage is collecting data, then selecting data. Selection of data is needed, if there is missing value data, the data will be discarded. After doing data cleansing, then the data is divided into two namely the training data and test data with the percentage of each 70% for training data and 30% for the test data. The training data consists of prerequisite value as a predictor variable and predetermined value as a response variable. Test data just as the training data contains some prerequisite and predetermined value.

We used a dataset extracted from the Petra Christian University’ Moodle. The main page of Petra Christian University’ Moodle can be seen in Figure 2. The dataset spans four semesters and it contains 486 courses, 7,563 students, and 109,231 activities. The courses belong to 21 different schools; each university student has registered in around 5 courses. In this research, the activities refer to the assignments and quizzes in Lentera. For each student-activity pair ( $s, a$ ), feature vector  $f_{sa}$  is constructed. There are three categories: student-centered features, activity-centered features and Lentera interaction features. Student-centered features are features related to the student. There are two categories:

- GPA\_total: The number of grades points a student earned in a given period of time.
- Grade\_total: The average grade accomplished over the entirety of the past exercise in the course.

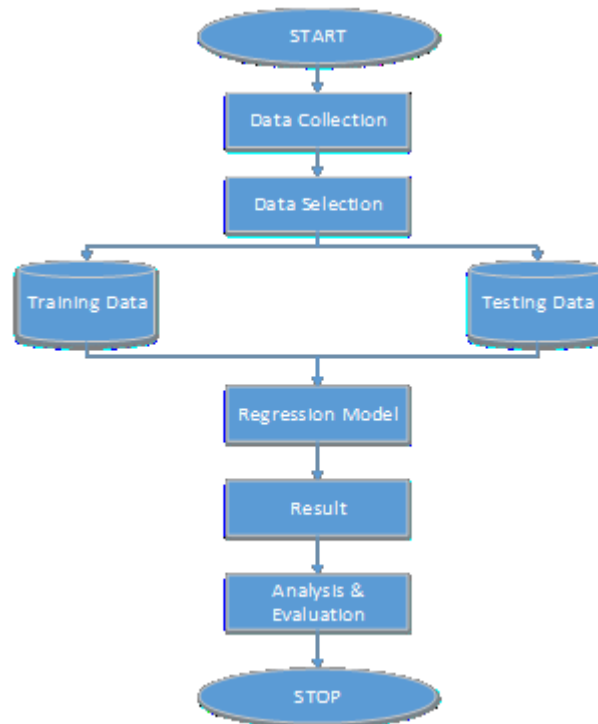


Figure 1. The Flow diagram of application design

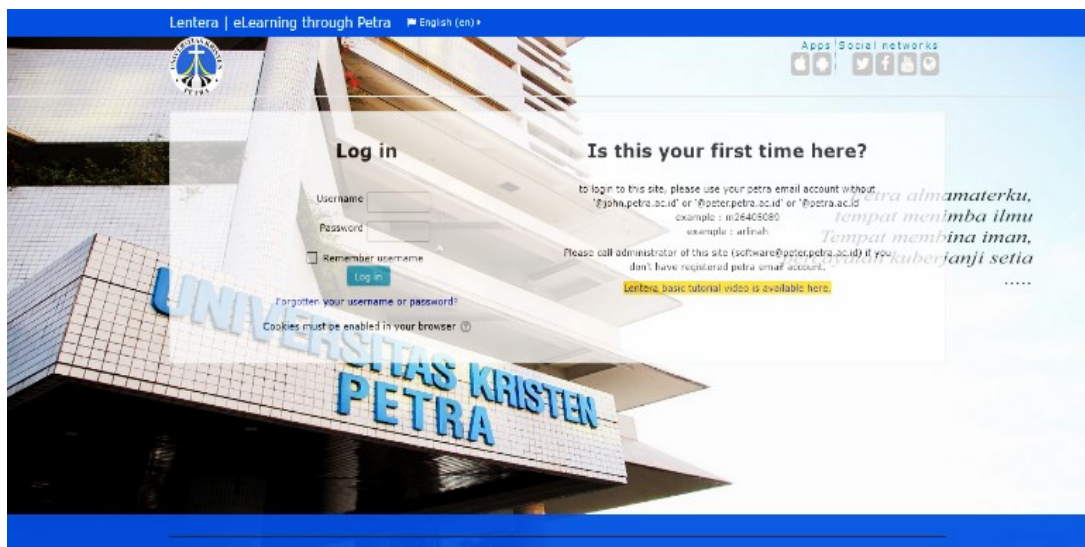


Figure 2. The main page of Lentera

Activity-centered features are features that relate to the activity of student in the Lentera LMS. Figure 3 describes the list of activities in Lentera. There are three categories:

- Activity\_type: activity of student in order to interact with other student or teacher in Lentera, This can either be quiz or assignment.
- Course\_level: The difficulty level of course. The range of value is 1, 2, 3 and 4. Value 1 means the difficulty of course is very low.
- Department: The department who offer the course.

Lentera-centered features describe the student's interaction with Lentera prior to the due date of the quizzes and assignments. These features were extracted from Lentera's log files and are the following:

- Discuss\_total: the number of discussion that posted by student.

- log\_total: frequency of the student login to the Lentera
- time\_total: total amount of time spent between login and logout
- read\_total: the number of discussions' forum that are delivered by the student.
- viewed\_total: the number of times the student viewed related material.

The dataset was divided into two subsets, namely training and testing subsets comprising 70% and 30% of the dataset respectively. The proposed model was trained on the training datasets and then evaluated on the testing datasets. This evaluation process was reiterated 5 times and the acquired results on the test datasets were calculated. The root mean squared error (RMSE) was used to assess the proposed model. It measures the difference between the actual and predicted grades on the test datasets.

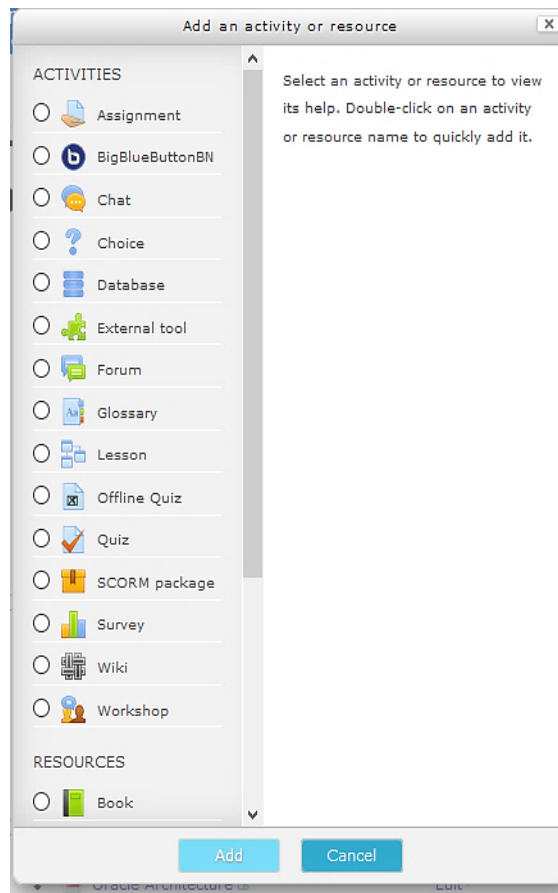


Figure 3. Activities in Lentera

### 3. RESEARCH FINDINGS AND ANALYSIS

This section presents the research findings and analysis. Moreover, the performance comparison between multi-regression model and single regression are discussed. Figure 4 shows the statistics in Lentera. It shows the number of active courses, students and activities in Lentera. The correlation between activities in Lentera (interaction between students with the Lentera features) and the predicted grades is discussed. To get the better result, the multi-regression models and the baseline model were trained 2 times.

Figure 5 shows the graphic of the single regression and the multi-regression models with and without using Lentera-interaction features. It can be seen from this figure, the value of RMSE was change along this experiments. It is clear from Figure 5 that the RMSE of multi regression model with Lentera features with one linear model is 0.17. On the other hand, the RMSE of single regression model is 0.3. By accompanying student-bias term and course-bias term, multi-regression model could better capture student performances in their course.

Figure 5 illustrates that there is a decrement of obtained RMSE by the multi regression model with increasing number of linear models. Using twelve proposed regression models, the acquired RMSE drops to 0.12. Comparing the performance of the two multi-regression models in Figure 5, we can see that the model

that uses the Lentera features performs better than the one that does not use them. A multi-regression model with ten linear models gives and an RMSE of 0.143 without using the Lentera features and gives an RMSE of 0.12 using the Lentera features. The use of Lentera features lead to more drop in RMSE with increasing number of regression models. From the evaluation, it can be concluded that it is because the proposed model that practices the Lentera features have extra student Lentera collaboration information to study from as the number of regression models increase.

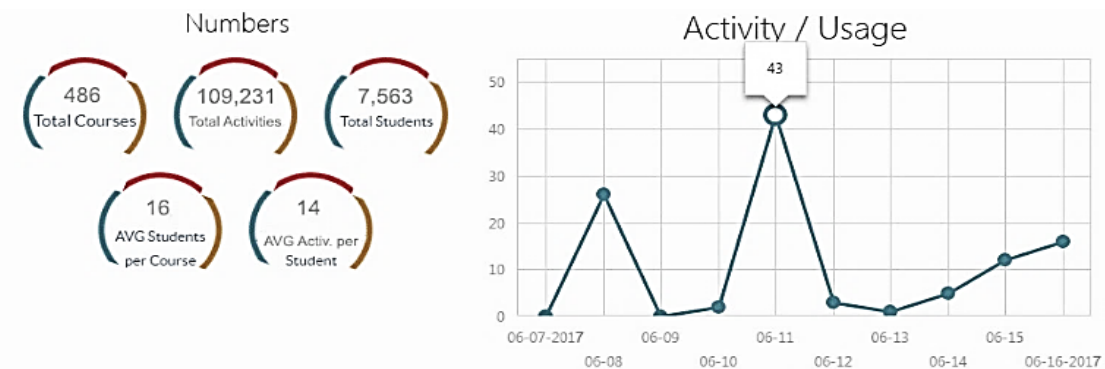


Figure 4. Statistics in Lentera

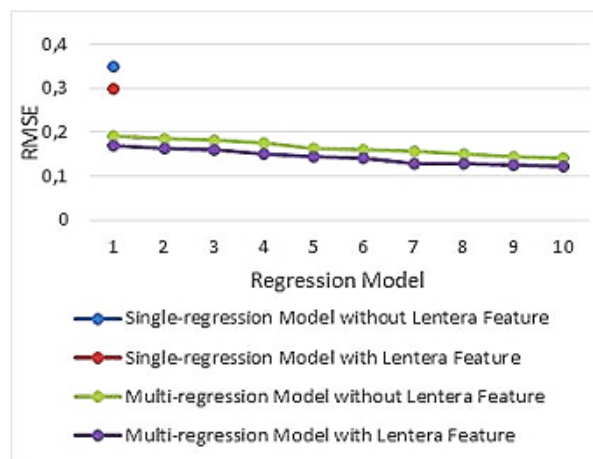


Figure 5. The graphic of regression model vs RMSE

#### 4. CONCLUSION

In this research, multi-regression model to forecast the performance of university student was implemented. According to the testing result, multi-regression model performs better in explaining dependent variables than single linear regression. Moreover, by increasing the number of linear regression model, the RMSE tends to decrease gradually. Finally, Lentera interaction features could improve the accuracy of prediction of student performance.

#### ACKNOWLEDGMENT

This research was supported by The Ministry of Research, Technology and Higher Education of the Republic of Indonesia. Research Grant Scheme (No: 002/SP2H/LT/K7/KM/2017).

#### REFERENCES

- [1] Santoso. L. W., "Analysis of the impact of information technology investments – a survey of Indonesian universities," *ARN JEAS*, vol. 9, no. 12, pp. 2404-2410, 2014.
- [2] Baker R. and Inventado. P., "Educational data mining and learning analytics," *Learning Analytics*, pp. 61-75, 2014.

- [3] Anderson J. R., Boyle C. F., and Reiser B. J., "Intelligent tutoring systems," *Science*, vol. 228, no. 4698, pp. 456-462, 1985.
- [4] Mjhool A. Y., Alhilali A. and H, Al-Augby S, "A Proposed architecture of big educational data using hadoop at the university of kufa," *International Journal of Electrical and Computer Engineering*, vol. 9, no. 6, pp. 4970-4978, 2019.
- [5] Romero C., and Ventura S., "Educational data mining: A review of the state of the art," *Trans. Sys. Man Cyber Part C*, vol. 40, no. 6, pp. 601-618, 2010.
- [6] Santoso L. W., Yulia, "The analysis of student performance using data mining," *Advances in Intelligent Systems and Computational Sciences*, pp. 559-573, 2019.
- [7] Santoso L. W., "Early warning system for academic using data mining," *Fourth International Conference on Advances in Computing, Communication & Automation*, pp. 1-4, 2019.
- [8] Santoso L. W., "Data warehouse with big data technology for higher education," *Procedia Computer Science*, vol. 124, no. 1, pp. 93-99, 2017.
- [9] Barber R., Sharkey M., "Course correction: Using analytics to predict course success," *International Conference on Learning Analytics and Knowledge*, pp. 259-262, 2012.
- [10] Wang J., and Karypis G., "Harmony: Efficiently mining the best rules for classification," *Data Mining Conference*, pp. 205-216, 2005.
- [11] Han J., Pei J., Yin Y., "Mining frequent patterns without candidate generation," *ACM SIGMOD Int'l Conf. on Management of Data*, vol. 29, no. 2, pp. 1-12, 2000.
- [12] Fradkin D. and Morchen F., "Mining sequential patterns for classification," *Knowl. Inf. Syst.*, vol. 45, no. 3, pp. 731-749, 2015.
- [13] Agrawal R., Golshan B., and Papalexakis E. E., "Toward data-driven design of educational courses: A feasibility Study," *Journal of Educational Data Mining (JEDM)*, vol. 8, no. 1, pp. 1-21, 2016.
- [14] Jittawiriyankoon C., "Proposed classification for elearning data analytics with MOA," *International Journal of Electrical and Computer Engineering*, vol. 9, no. 5, pp. 3569-3575, 2019.
- [15] Andayani S., et al, "Decision-making model for student assessment by unifying numerical and linguistic data," *International Journal of Electrical and Computer Engineering*, vol. 7, no. 1, pp. 363-373, 2017.
- [16] Santoso L. W., "ITIL service management model for e-learning," *Journal of Adv Research in Dynamical & Control Systems*, vol. 11, no. 6, pp. 190-197, 2019.
- [17] Quadri M. N., and Kalyankar N. V., "Drop out feature of student data for academic performance using decision tree techniques," *Glob. J. Comput. Sci. Technology*, vol. 10, no. 2, pp. 2-5, 2010.
- [18] Dekker G. W., Pechenizkiy M., and Vleeshouwers J. M., "Prediction student drop out: A case study," *2nd International Conference on Educational Data Mining*, pp. 41-50, 2009.
- [19] Barber R., and Sharkey M., "Course correction: Using analytics to predict course success," *2nd International Conference on Learning Analytics and Knowledge*, pp. 259-262, 2012.
- [20] Leitner P., Khalil M., and Ebner M., "Learning analytics in higher education δ a literature review," *Learning Analytics: Fundaments, Applications, and Trends. Springer International Publishing*, pp. 1-23, 2017.
- [21] Lee Y., and Cho J., "An intelligent course recommendation system," *Smart CR*, vol. 1, no. 1, pp. 69-84, 2011.
- [22] Yunianta A., et al., "Solving the complexity of heterogeneity data on learning environment using ontology," *TELKOMNIKA Telecommunication Computing Electronics and Control*, vol. 13, no. 1, pp. 341-348, 2015,
- [23] Ventura C. S., Espejo P. G., and Hervás C., "Data mining algorithms to classify students," *1st Int. Conf. on Educational Data Mining. Montreal*, pp. 187-191, 2008.
- [24] Thai-Nghe Ng, et al., "Factorization techniques for predicting student performance. educational recommender systems and technologies: Practices and challenges," *OIGI Global*, pp. 129-153, 2011.
- [25] Superby J. F, Vandamme J. P., and Meskens N, "Determination of factors influencing the achievement of the first-year university students using data mining methods," *Workshop on educational data mining*, vol. 32, 2006.
- [26] Lamani A., et al., "Data mining techniques application for prediction in OLAP Cube," *International Journal of Electrical and Computer Engineering*, vol. 9, no. 3, pp. 2094-2102, 2019.
- [27] Daniel B., "Big data and analytics in higher education: Opportunities and challenges," *British Journal of Educational Technology*, vol. 46, no. 5, pp. 904-920, 2014.
- [28] Elbadrawy A., Studham R. S., and Karypis G., "Personalized multi-regression models for predicting students' performance in course activities," *International Conference on Learning Analytics and Knowledge*, pp. 103-107, 2015.
- [29] Agarwal D., and Chen B. C., "Regression-based latent factor models," *ACM SIGKDD Conference on Knowledge Discovery and Data Mining*, pp. 19-27, 2009.