

4th ICITB

K-NEAREST NEIGHBOR METHOD FOR MONITORING OF PRODUCTION AND PRESERVATION INFORMATION (TREATMENT) OF RUBBER TREE PLANT

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ABSTRACT

The District in Karang Agung Village, the report of production results by the foreman is still done by making regular reports in the foreman's order, supervision of production by the assistant head of the region is still done by looking at reports from the foreman who is still in the form of regular records in the foreman's book and there are still many rubber plants that are not well developed due to lack of knowledge about how to preserve the rubber plants. The methods used have weaknesses, including foremen having difficulty reporting production data because of the distance of plantations and remote plantation offices. In addition, the assistant head of the region was able to find it difficult to monitor production results because the data provided was still in the form of records and is often late by the foreman. The algorithm analysis used the classification method, K-Nearest Neighbor algorithm. A classification method for a set of data based on previously classified data learning. The nearest Neighbor was an approach to finding a case by calculating the proximity between a new case and an old case. It was based on matching the weights of a number of features.

Keywords: District, production, K-Nearest Neighbor, Classification

INTRODUCTION

In the District in Karang Agung Village, the foreman's production report is still done by making regular reports for the foreman, supervision of production by the assistant head of the region is also still done by looking at the foreman's report which is still a regular record in the foreman's book and there are still many rubber plants that are not well developed due to lack of knowledge about how to preserve (care) the rubber plants. The method used has a weakness, for example, the foreman has difficulty reporting production data due to the distance of plantations and remote plantation offices, besides that assistant head of the region will find it difficult to monitor production results because the data provided is still in the form of notes and is often late given by the foreman. And the difficulty in accessing information if you are still searching manually on a website one to another website. Based on this, a system is needed that can input data on rubber latex production by the foreman, and the assistant head of the region can oversee the production of rubber plants. And the media needed to make it easier to access information about rubber plants. For this reason, an application for monitoring the results of production and information on preservation (maintenance) of rubber plants is built using the K-Nearest Neighbor method which is expected to help the performance in PTP Nusantara Vii in Karang Agung Village Unit.

LITERATURE REVIEW

Rubber plants (*Hevea Brasiliensi*) are important plantation crops both in the economic context of the community and non-oil and gas foreign exchange sources for the country. Rubber plants come from tropical regions of the Amazon Brazillia with rainfall from 2000 to 3000 mm/year and rainy days between 120-170 days/year [1]. Development of concentrated rubber in areas 10 LU and 10 LS Most areas of Indonesian rubber plantations are located in Sumatra (70%), Kalimantan (24%) and Java (4%) with rainfall of 1500-4000 mm / year with average dry months 0-4 months per year and is located at an elevation below 500 m above sea level. [1]

K-Nearest Neighbor (KNN) Algorithm

KNN is done by finding groups of objects in the closest (similar) training data to objects on new data or testing data. K-Nearest Neighbor Algorithm is a method for classifying objects based on learning data which is the closest distance to the object. Nearest Neighbor is an approach to look for cases by calculating the closeness between new cases and old cases which is based on matching weights of a number of features. The K-NN classification works based on an analogy, where test and training data are compiled and drawn conclusions based on the similarity of the data produced by the comparison.

$$d_{ik} = \sqrt{\sum_j^m (C_{ij} - C_{kj})^2}$$

Figure 1. Formula of Knn Algorithm

Where :

C_{ij} = Sample data

C_{kj} = New data

K-Nearest Neighbor (KNN) is a method for classifying new objects based on (K) their closest neighbors. KNN includes a supervised learning algorithm, where the results of new query instances are classified based on the majority of the categories on the KNN. The most appearing classes will be the class of classification. [6]

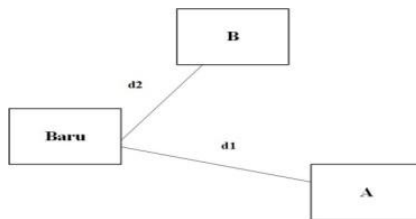


Figure 2. Illustration of K-Nearest Neighbor Algorithm

features. To find which cases to use, the closeness between new cases and all old cases is calculated. The old case with the greatest closeness will be the solution to be used in new cases. [7]

RESEARCH METHOD

Analysis of Algorithms used The steps of the K-Nearest Neighbor algorithm are as follows:

- a) Determine the parameter k (number of closest neighbors).
- b) Calculates the square of the Euclidean distance of an object against the training data provided
- c) Sort results in number 2 in sequence.
- d) Collect Y categories (The nearest neighbor classification is based on the value of k)
- e) By using the most majority nearest neighbor category, object categories can be predicted.

Application Design (System And Software Design)

Application design is the stage of system design or application which includes the preparation of processes, data, process flow, and fulfillment of needs in accordance with the results of the needs analysis. Application

design documentation generated from this stage is a use case diagram and activity diagram. [8]

UML design

Draft the proposed application case diagram.

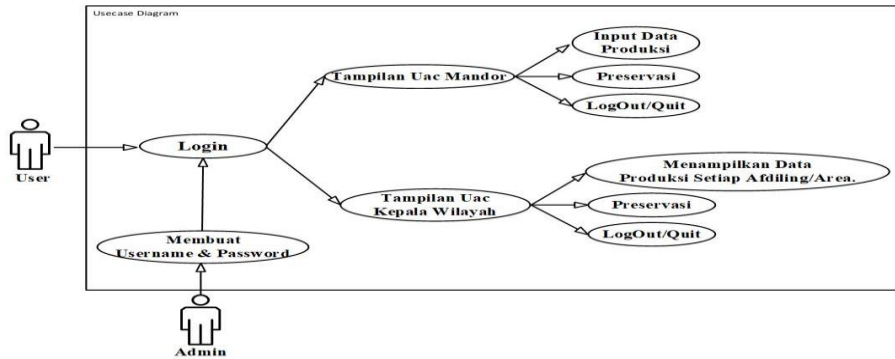


Figure 3. Use Case Proposed Application Chart

FINDINGS AND DISCUSSION

Preservation Page Views

On the preservation page it displays articles about preservation of rubber plants, and this preservation page can be accessed by the assistant head of region. Preservation page views can be seen in Figure 4.

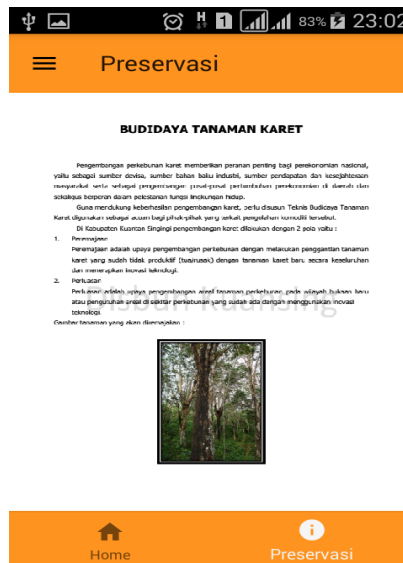


Figure 4. (Preservation Page Views)

Discussion of K-Nearest Neighbor Algorithm

K-Nearest Neighbor (K-NN) algorithm is a method of classification of a set of data based on data learning that has been previously classified. Included in supervised learning, where the results of querying new instances are classified based on the majority of the proximity of the categories in the K-NN. The following is a table of production results at PTP. Nusantara VII unit of Karang Agung Village from January to December 2017.

Table 1. Production Data

| Area | Production Results (Kg) | | | | | | | | | | | |
|----------|-------------------------|----------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|
| | Januari | Februari | Maret | April | Mei | Juni | Juli | Agustus | Sept | Okt | Nov | Des |
| AFD I | 119.912 | 120.216 | 149.291 | 150.221 | 151.967 | 161.755 | 154.211 | 136.132 | 113.194 | 120.125 | 132.321 | 150.112 |
| AFD II | 100.512 | 101.912 | 135.998 | 137.210 | 141.003 | 142.725 | 138.159 | 124.250 | 94.239 | 101.211 | 122.150 | 137.210 |
| AFD III | 108.214 | 110.112 | 139.514 | 140.291 | 139.980 | 142.252 | 139.255 | 126.109 | 96.112 | 109.104 | 124.291 | 140.191 |
| AFD IV | 124.712 | 124.121 | 125.100 | 126.161 | 130.209 | 151.202 | 143.108 | 138.185 | 114.397 | 125.791 | 135.129 | 125.721 |
| AFD V | 114.292 | 113.891 | 172.001 | 175.161 | 175.999 | 172.407 | 163.204 | 133.176 | 101.008 | 115.921 | 130.192 | 172.216 |
| AFD VI | 110.912 | 111.132 | 147.412 | 150.126 | 155.461 | 161.222 | 159.301 | 124.341 | 98.143 | 110.213 | 124.912 | 146.421 |
| AFD VII | 121.319 | 120.192 | 163.555 | 163.692 | 109.300 | 170.125 | 165.112 | 134.102 | 104.107 | 121.111 | 133.291 | 164.912 |
| AFD VIII | 100.912 | 101.219 | 132.191 | 135.729 | 136.012 | 135.001 | 130.921 | 114.108 | 86.231 | 100.012 | 115.119 | 132.191 |

Production data from January-December 2017 are 96 data. Where in the 2017 production data, the lowest production results were found in the AFD II area in September with the production yield of 86,231 Kg and the highest yield was found in the AFD V area in May with a yield of 175,999 Kg. Classification of production is divided into 3 categories, namely:

- a) Low production category: With production yield of $\leq 120,215$ Kg
- b) Medium production category: With production results of 20120,216 Kg up to 140,290 Kg
- c) Category of high production yield: With production yield $\geq 140,300$ Kg

Calculate the distance of new data with training data for each Area using the Euclidean distance formula

$$d_{ik} = \sqrt{\sum_i^m (C_{ij} - C_{kj})^2}$$

Information;

dik = distance

Cij = training data / sample

Ckj = New data

Manual calculation uses the K-nearest Neighbor algorithm with new data or data testing = 100,000 Kg with K = 4.

Table 2. Production Data Calculation Tables

| No | Production Data | Calculate Euclidean distance | Smallest Distance | K=4 | Category |
|----|-----------------|--------------------------------|-------------------|-----|----------|
| 1 | 86.231 | $(86.231-100.000)^2 = 13.769$ | 20 | No | Low |
| 2 | 94.239 | $(94.239-100.000)^2 = 5.761$ | 11 | No | Low |
| 3 | 96.112 | $(96.112-100.000)^2 = 3.888$ | 9 | No | Low |
| 4 | 98.143 | $(98.143-100.000)^2 = 1.857$ | 8 | No | Low |
| 5 | 100.012 | $(100.012-100.000)^2 = 12$ | 1 | Yes | Low |
| 6 | 100.512 | $(100.512-100.000)^2 = 512$ | 2 | Yes | Low |
| 7 | 100.912 | $(100.912-100.000)^2 = 912$ | 3 | Yes | Low |
| 8 | 101.008 | $(101.008-100.000)^2 = 1.008$ | 4 | Yes | Low |
| 9 | 101.211 | $(101.211-100.000)^2 = 1.211$ | 5 | No | Low |
| 10 | 101.219 | $(101.219-100.000)^2 = 1.219$ | 6 | No | Low |
| 11 | 101.219 | $(101.219-100.000)^2 = 1.219$ | 7 | No | Low |
| 12 | 104.107 | $(104.107-100.000)^2 = 4.107$ | 10 | No | Low |
| 13 | 108.214 | $(108.214-100.000)^2 = 8.214$ | 12 | No | Low |
| 14 | 109.104 | $(109.104-100.000)^2 = 9.104$ | 13 | No | Low |
| 15 | 109.300 | $(109.300-100.000)^2 = 9.300$ | 14 | No | Low |
| 16 | 110.112 | $(110.112-100.000)^2 = 10.112$ | 15 | No | Low |
| 17 | 110.213 | $(110.213-100.000)^2 = 10.213$ | 16 | No | Low |
| 18 | 110.912 | $(110.912-100.000)^2 = 10.912$ | 17 | No | Low |

| | | | | | |
|----|---------|-----------------------------------|----|----|--------|
| 19 | 111.132 | $(111.132-100.000)^2$ = 11.132 | 18 | No | Low |
| 20 | 113.194 | $(113.194-100.000)^2$ = 13.194 | 19 | No | Low |
| 21 | 113.891 | $(113.891-100.000)^2$ = 13.891 | 21 | No | Low |
| 22 | 114.108 | $(114.108-100.000)^2$ = 14.108 | 22 | No | Low |
| 23 | 114.292 | $(114.292-100.000)^2$ = 14.292 | 23 | No | Low |
| 24 | 114.397 | $(114.297-100.000)^2$ = 14.297 | 24 | No | Low |
| 25 | 115.119 | $(115.119-100.000)^2$ = 15.119 | 25 | No | Low |
| 26 | 115.921 | $(115.921-100.000)^2$ = 15.921 | 26 | No | Low |
| 27 | 119.912 | $(119.912-100.000)^2$ = 19.912 | 27 | No | Low |
| 28 | 120.125 | $(119.912-100.000)^2$ = 19.912 | 28 | No | Low |
| 29 | 120.192 | $(120.192-100.000)^2$ =20.192 | 29 | No | Low |
| 30 | 120.216 | $(120.216-100.000)^2$ = 20.216 | 30 | No | Low |
| 31 | 121.111 | $(121.111-100.000)^2$ = 21.111 | 31 | No | Medium |
| 32 | 121.319 | $(121.319-100.000)^2$ = 21.319 | 32 | No | Medium |
| 33 | 122.150 | $(122.150-100.000)^2$ = 22.150 | 33 | No | Medium |
| 34 | 124.121 | $(124.121-100.000)^2$ = 24.121 | 34 | No | Medium |
| 35 | 124.250 | $(124.250-100.000)^2$ = 24.250 | 35 | No | Medium |
| 36 | 124.291 | $(124.291-100.000)^2$ | 36 | No | Medium |

| | | | | | |
|----|---------|-----------------------------------|----|----|--------|
| | | = 24.291 | | | |
| 37 | 124.341 | $(124.341-100.000)^2$ = 24.341 | 37 | No | Medium |
| 38 | 124.712 | $(124.712-100.000)^2$ = 24.712 | 38 | No | Medium |
| 39 | 124.912 | $(124.912-100.000)^2$ = 24.912 | 39 | No | Medium |
| 41 | 125.100 | $(125.100-100.000)^2$ = 25.100 | 40 | No | Medium |
| 41 | 125.721 | $(125.721-100.000)^2$ = 25.721 | 41 | No | Medium |
| 42 | 125.791 | $(125.791-100.000)^2$ = 25.791 | 42 | No | Medium |
| 43 | 126.109 | $(126.109-100.000)^2$ = 26.109 | 43 | No | Medium |
| 44 | 126.161 | $(126.161-100.000)^2$ = 26.161 | 44 | No | Medium |
| 45 | 130.192 | $(130.192-100.000)^2$ = 30.192 | 45 | No | Medium |
| 46 | 130.209 | $(130.209-100.000)^2$ = 30.209 | 46 | No | Medium |
| 47 | 130.921 | $(130.921-100.000)^2$ = 30.921 | 47 | No | Medium |
| 48 | 132.191 | $(132.191-100.000)^2$ = 32.191 | 48 | No | Medium |
| 49 | 132.191 | $(132.191-100.000)^2$ = 32.191 | 49 | No | Medium |
| 50 | 132.321 | $(132.321-100.000)^2$ = 32.321 | 50 | No | Medium |
| 51 | 133.176 | $(133.176-100.000)^2$ = 33.176 | 51 | No | Medium |
| 52 | 133.291 | $(133.291-100.000)^2$ = 33.291 | 52 | No | Medium |
| 53 | 134.102 | $(134.102-100.000)^2$ = 34.102 | 53 | No | Medium |

| | | | | | |
|----|---------|-----------------------------------|----|----|--------|
| 54 | 135.001 | $(135.001-100.000)^2$ = 35.001 | 54 | No | Medium |
| 55 | 135.129 | $(135.129-100.000)^2$ = 35.129 | 55 | No | Medium |
| 56 | 135.729 | $(135.729-100.000)^2$ = 35.729 | 56 | No | Medium |
| 57 | 135.998 | $(135.998-100.000)^2$ = 35.998 | 57 | No | Medium |
| 58 | 136.012 | $(136.012-100.000)^2$ = 36.012 | 58 | No | Medium |
| 59 | 136.132 | $(136.132-100.000)^2$ = 36.132 | 59 | No | Medium |
| 60 | 137.210 | $(137.210-100.000)^2$ = 37.210 | 60 | No | Medium |
| 61 | 137.210 | $(137.210-100.000)^2$ = 37.210 | 61 | No | Medium |
| 62 | 138.159 | $(138.159-100.000)^2$ = 38.159 | 62 | No | Medium |
| 63 | 138.185 | $(138.185-100.000)^2$ = 38.185 | 63 | No | Medium |
| 64 | 139.255 | $(139.255-100.000)^2$ = 39.255 | 64 | No | Medium |
| 65 | 139.514 | $(139.514-100.000)^2$ = 39.514 | 65 | No | Medium |
| 66 | 139.980 | $(139.980-100.000)^2$ = 39.980 | 66 | No | Medium |
| 67 | 140.191 | $(140.191-100.000)^2$ = 40.191 | 67 | No | High |
| 68 | 140.291 | $(140.291-100.000)^2$ = 40.291 | 68 | No | High |
| 69 | 141.003 | $(141.003-100.000)^2$ = 41.003 | 69 | No | High |
| 70 | 142.252 | $(142.252-100.000)^2$ = 42.252 | 70 | No | High |
| 71 | 142.725 | $(142.725-100.000)^2$ | 71 | No | High |

| | | | | | |
|----|---------|-----------------------------------|----|----|------|
| | | = 42.725 | | | |
| 72 | 143.108 | $(143.108-100.000)^2$ = 43.108 | 72 | No | High |
| 73 | 146.421 | $(146.421-100.000)^2$ = 46.421 | 73 | No | High |
| 74 | 147.412 | $(147.412-100.000)^2$ = 47.412 | 74 | No | High |
| 75 | 149.291 | $(149.291-100.000)^2$ = 49.291 | 75 | No | High |
| 76 | 150.112 | $(150.112-100.000)^2$ = 50.112 | 76 | No | High |
| 77 | 150.126 | $(150.126-100.000)^2$ = 50.126 | 77 | No | High |
| 78 | 150.221 | $(150.221-100.000)^2$ = 50.221 | 78 | No | High |
| 79 | 151.202 | $(151.202-100.000)^2$ = 51.202 | 79 | No | High |
| 80 | 151.967 | $(151.967-100.000)^2$ = 51.967 | 80 | No | High |
| 81 | 154.211 | $(154.211-100.000)^2$ = 54.211 | 81 | No | High |
| 82 | 155.461 | $(155.461-100.000)^2$ = 55.461 | 82 | No | High |
| 83 | 159.301 | $(159.301-100.000)^2$ = 59.301 | 83 | No | High |
| 84 | 161.222 | $(161.222-100.000)^2$ = 61.222 | 84 | No | High |
| 85 | 161.755 | $(161.755-100.000)^2$ = 61.755 | 85 | No | High |
| 86 | 163.204 | $(163.204-100.000)^2$ = 63.204 | 86 | No | High |
| 87 | 163.555 | $(163.555-100.000)^2$ = 63.555 | 87 | No | High |
| 88 | 163.692 | $(163.692-100.000)^2$ = 63.692 | 88 | No | High |

| | | | | | |
|----|---------|-----------------------------------|----|----|------|
| 89 | 164.912 | $(164.912-100.000)^2$ = 64.912 | 89 | No | High |
| 90 | 165.112 | $(165.112-100.000)^2$ = 65.112 | 90 | No | High |
| 91 | 170.125 | $(170.125-100.000)^2$ = 70.125 | 91 | No | High |
| 92 | 172.001 | $(172.001-100.000)^2$ = 72.001 | 92 | No | High |
| 93 | 172.216 | $(172.216-100.000)^2$ = 72.216 | 93 | No | High |
| 94 | 172.407 | $(172.407-100.000)^2$ = 72.407 | 94 | No | High |
| 95 | 175.161 | $(175.161-100.000)^2$ = 75.161 | 96 | No | High |
| 96 | 172.999 | $(172.999-100.000)^2$ = 72.999 | 95 | No | High |


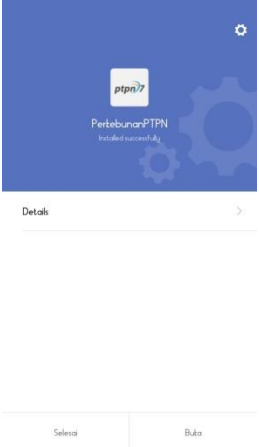
By sorting the smallest distance, $K = 4$ is taken, and the 4 closest distance is included in the low category, it can be concluded that the production rate of 100,000 Kg includes low production yield.

Software System Testing

Installation Testing

Installation testing is done whether the built-in application can run on the Android operating system.

Table 3. Installation testing

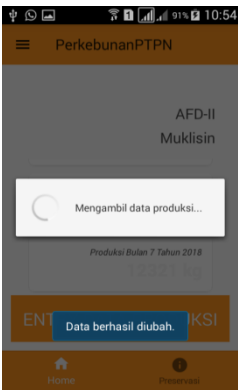

| No. | Brand | Specification | Test result | Information |
|-----|-------------------|--|---|-------------|
| 1. | Samsung J1 Ace | RAM 0.75GB Android 4.4.4 KitKat Speed HSPA 21.1/5.76 Mbps, LTE Ca t4 150/50 Mbps. Ukuran Layar 4.3 inches |  | SUCCESS |
| 2 | Xiomi Redmi 4x | |  | SUCCESS |

Usage Testing

At the stage of testing the use, will be tested on several smartphones as

follows: The results of the test can be seen in table 4.

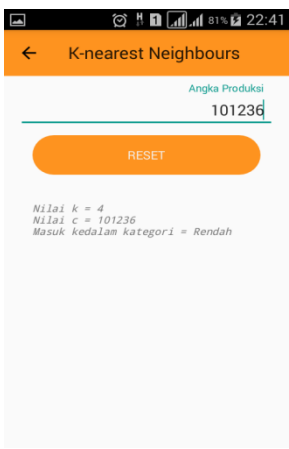
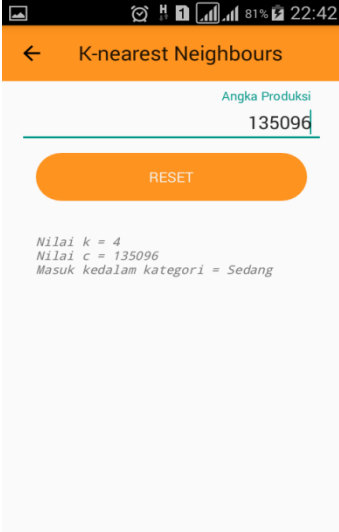
Table 4. Testing on inputting production data


| No. | Brand | Specification | Test result | Information |
|-----|-------------------|---|--|-------------|
| 1. | Samsung J1 Ace | RAM 0.75GB |  | SUCCESS |
| | | Android 4.4.4 KitKat | | |
| | | Speed HSPA 21.1/5.76 Mbps, LTE Ca t4 150/50 Mbps. | | |
| | | Ukuran Layar 4.3 inches | | |
| 2 | Xiomi Redmi 4x | |  | SUCCESS |
| | | | | |
| | | | | |
| | | | | |

Testing of K-Nearest Neighbor Algorithm

K-Nearest Neighbor software testing can be seen in table 5.

Table 5. Software algorithms

| No | Brand | Specification | Test result | Information |
|----|----------------|---|--|--|
| 1 | Samsung J1 Ace | RAM 0.75GB Android 4.4.4 KitKat Speed HSPA 21.1/5.76 Mbps, LTE Cat 4 150/50 Mbps. |  | New data is included in the low category based on the calculation of the algorithm by finding the closest distance to the old data that has been classified into the low, medium and high categories. |
| 2 | Samsung J1 Ace | RAM 0.75GB |  | New data is included in the medium category based on the calculation of the algorithm by finding the closest distance to the old data that has been classified into the low, medium and high categories. |

| | | | | |
|---|-------------------|------------|--|--|
| 2 | Samsung J1 Ace | RAM 0.75GB |  | New data is included in the High category based on the calculation of the algorithm by finding the closest distance to the old data that has been classified into the low, medium and high categories. |
|---|-------------------|------------|--|--|

CONCLUSIONS

Using the K-Nearest Neighbor method can help performance at PTP Nusantara VII in Karang Agung Village Unit. This system is proposed to make it easier for the foreman and assistant head of region and foreman to no longer carry out conventional/procedural manuals and can store the database as a production data archive.

REFERENCES

- [1] Sabarman Damanik. 2012. Pengembangan Karet (*Havea Brasiliensis*) Berkelanjutan di Indonesia. Bogor : Perspektif Vol.11 No.1 Hlm 91-102 ISSN:1412-8004.
- [2] Nazruddin, S. H. (2015). Pemrograman Aplikasi Mobile Smartphone dan Tablet Pc Berbasis Android. Bandung: Informatika
- [3] Mudjahidin & Nyoman Dita Pahang Putra. 2010. Rancang Bangun Sistem Informasi Monitoring Perkembangan Proyek Berbasis Web Studi Kasus di Dinas Bina Marga dan Pemantusan. Surabaya : Jurnal Teknik Industri, Vol. 11, No. 1, Febuari 2010: 75-83.
- [4] Moh. Rochman Wahid Maulana. 2017. Pengembangan Aplikasi Untuk Studi Bahasa Carakan Madura : PROSIDING ISSN: 2549 - 869X.

- [5] Angga Setiyadi & Tati Hariyanti. 2016. Penerapan Sqlite Pada Aplikasi Pengaturan Waktu Ujian dan Presentasi. Bandung : Jurnal Teknik Informatika, Vol.13 No.2.
- [6] Nuraini, R. (2015). Desain Algoritma Perkalian Matriks Menggunakan Metode Flowchart. *Jurnal Teknik Komputer Amik BSI*, 144-151.
- [7] Toto Andri Puspito. 2017. Menentukan Pilihan Sekolah Didalam Penerimaan Peserta Didik Baru Dengan Menggunakan Metode Nave bayes dan K-Nearest Neighbor (Studi Kasus: PPDB Online Jenjang SMP Kota Metro). Bandar Lampung : Jurnal Informatika, Vol. 17, No. 1.
- [8] Roger S.Pressman, Ph.D. (2012). *Rekayasa Perangkat Lunak (pendekatan praktisi edisi 7)*. Yogyakarta: penerbit C.V ANDI.