# INTELLIGENT LEARNING ACHIEVEMENT PREDICTION SYSTEM USING K-MEANS ALGORITHM AT UPT EDUCATION UNIT SMPN 5 BLITAR

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#### Abstract

This research aims to develop an intelligent system that can predict student performance using the K-Means clustering algorithm in the UPT teaching unit SMPN 5 Blitar. In data collection methods, student scores are taken as input into intelligent systems by educators teaching these subjects. This kind of research is done using quantitative research. The study collected data from a sample of a total of 227 students, with student performance outcomes as input variables. The K-means clustering method is used to classify students into subject grades: subject grades (Religious Education, Indonesian Language, English, Mathematics, Science, Social Studies, Cultural Studies, PJOK, and Computer Science) and behavioral scores (attitude and discipline). After testing using the Silhouette Coefficients data test, the accuracy of the data reached 86%. The conclusion of this study is that the implementation of the K-Means clustering method improves the efficiency and accuracy of identifying bright students.

**Keywords**: Intelligent System, Learning Achievement Prediction, K-Means clustering algorithm, Clustering, silhouette coefficient.

## 1. INTRODUCTION

One of the public secondary education institutions in Blitar City, namely the UPT Education Unit of SMPN 5 Blitar, is a secondary school that has a vision to become an advanced school and become a competitive force and interest by students in Blitar City. Educators at UPT Education Unit SMPN 5 Blitar mostly still use academic grades as a benchmark to see the level of achievement of their students. Every year the teacher collects a list of grades using the ms excel application. After that, at the end of the semester, the total score of all students will be added up, from here the teachers at the UPT Education Unit of SMPN 5 Blitar can determine the outstanding students. If the grades have all been processed, they will be deposited into the curriculum to find out whether the student is outstanding or not. So that making the assessment takes a long time to determine outstanding students. The assessment is deemed ineffective and inefficient.

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Based on existing problems, therefore using the application of Data Mining to make it easier for the school curriculum to select students to enter superior classes, this study uses the K-Means Algorithm Clustering Method to determine the results of cluster grouping of outstanding students because this algorithm is an algorithm that has a fairly high accuracy on the size of the object, so this algorithm is relatively more scalable and efficient for processing large numbers of objects.

There are several previous studies that have been conducted by other researchers regarding superior classes but using other algorithms such as in research (Hutasuhut, 2019) with the Iterative Dichotomiser 3 (ID3) algorithm, then in research using the K-Means Clustering algorithm and the same dataset, namely superior classes, but using a computerized Academic Information System in research (Kusuma, 2019). While research using the same algorithm but with different datasets used such as the K-Means Clustering algorithm in research (Sadewo, 2017). K-Means Clustering algorithm in research (Yulianti, 2019). Fuzzy C-Means and K-Means algorithms in research (Butarbutar, 2016). And the K-Means Clustering Algorithm in research (Praja, 2019). So based on the background above, the author is interested in making a research entitled "Implementation of the K-Means Algorithm to Determine Superior Classes Based on Student Grades at SMPN 1 BOJONG".

Based on the above problems, a research title was made about "CERDASY SYSTEM PREDICTION OF STUDENT PRACTICE USING K-MEANS ALGORITHMA ON CLASS 7 STUDENTS IN UPT SATUAN EDUCATION SMPN 5 BLITAR". So it can be identified into several problems, namely that the teacher only relies on one academic score to find out which students are outstanding and the lack of efficiency of student value data processed using the Ms. Excel application.

# 2. METHOD

The research location was at SMPN 5 Blitar from February to August 2023. For the type of research used in this study is quantitative research in which the main data used is obtained from the recap of grades from all teachers who teach in grades 7A to 7H in odd and even semesters at UPT Education Unit SMPN 5 Blitar City.

For data collection techniques, researchers use interview techniques, namely by conducting observations, interviews and literature studies. Meanwhile, the type of data used is primary data, namely the original source or first party.

The data collection instrument uses data collection and problem solving, namely from the teaching teacher of each class subject 7A to 7H. For the first stage of the first research is by collecting data, data cleaning, data transformation (attitude and discipline value data), testing K-Means Clustering data. After calculating K-Means Clustering in determining the superior class, it is obtained that the group of outstanding students (Cluster 2) and the group of non-achieving students (Cluster 1) with reference to entering grades from Religion, Indonesian Language, English, Mathematics, Science, Social Studies, Cultural Arts, PJOK, Informatics, attitude and discipline scores. The value is obtained from the recap of the student teacher who teaches in class 7A to class 7H.

Furthermore, the first Silhouette Coefficient test carried out is the calculation of a(i). a(i) is to find the average distance between points in one cluster. Next, the Silhouette Coefficient test is carried out by calculating b(i). b(i) is to find the average distance between points in the overall data.

From the results of Silhouette Coefficient K-Means clustering and each distance calculation method, the results of each distance are obtained, after obtaining the results of s(i), the average value will be taken and used as the Silhouette Coefficient value. The average result of the calculation of the average distance in one cluster (a(i)) and the average distance between all points (b(i)) obtained the result s(i) = 1 which means that object i is in the right cluster, which means the clustering results are correct.

# 3. RESULTS AND DISCUSSION

The data used in this study is a collection of student values from class 7A to class 7H obtained from the subject teacher in charge of the class totaling 227 students. The following is the data that has been obtained.

| No  | Nama Siswa                   | Kelas | Agama | PKN | Indo | Ing | Mat | IPA | IPS | Seni | pjok | Info<br>r | sika<br>p | disip<br>lin |
|-----|------------------------------|-------|-------|-----|------|-----|-----|-----|-----|------|------|-----------|-----------|--------------|
| 1.  | Adelia Wahyu Setyoningsih    | 7A    | 81    | 81  | 81   | 80  | 83  | 80  | 80  | 81   | 82   | 84        | В         | В            |
| 2.  | Aji Eka Saputra              | 7A    | 84    | 79  | 79   | 77  | 78  | 79  | 77  | 79   | 85   | 83        | D         | А            |
| 3.  | Andi Adrian                  | 7A    | 79    | 80  | 80   | 78  | 78  | 77  | 78  | 80   | 85   | 80        | С         | А            |
| 4.  | Andrean Titan Ridho Pangestu | 7A    | 92    | 95  | 92   | 97  | 93  | 90  | 93  | 94   | 95   | 92        | А         | А            |
| 5.  | Anggun Gunawan               | 7A    | 85    | 84  | 84   | 85  | 83  | 82  | 80  | 84   | 84   | 86        | В         | А            |
| 6.  | Armando Junian Nugroho       | 7A    | 93    | 94  | 95   | 92  | 97  | 82  | 78  | 84   | 83   | 94        | А         | А            |
| 7.  | Asyraf Pratama               | 7A    | 81    | 82  | 92   | 95  | 92  | 97  | 78  | 82   | 85   | 92        | А         | А            |
| 8.  | Ayumi Talitha Bernadine      | 7A    | 82    | 78  | 77   | 76  | 77  | 76  | 76  | 79   | 82   | 79        | D         | D            |
| 9.  | Bagus Permadi                | 7A    | 80    | 81  | 80   | 77  | 78  | 78  | 78  | 81   | 82   | 83        | С         | В            |
| 10. | Cinta Bella Oktavia          | 7A    | 85    | 84  | 84   | 85  | 83  | 82  | 80  | 84   | 84   | 86        | В         | А            |
|     |                              |       |       |     |      |     |     |     |     |      |      |           |           |              |
| 223 | Nilam Rhamadani              | 7H    | 80    | 80  | 81   | 77  | 78  | 66  | 77  | 82   | 82   | 81        | В         | В            |
| 224 | Revalina Hafiz Azalia        | 7H    | 80    | 83  | 84   | 78  | 78  | 79  | 78  | 82   | 82   | 83        | С         | А            |
| 225 | Rivalno Septiansah           | 7H    | 82    | 80  | 80   | 78  | 79  | 80  | 78  | 82   | 81   | 82        | С         | А            |
| 226 | Satria Briant Ardinata       | 7H    | 79    | 78  | 77   | 66  | 76  | 76  | 76  | 78   | 83   | 79        | В         | В            |
| 227 | Savira Putri Felisha         | 7H    | 83    | 82  | 84   | 79  | 81  | 81  | 79  | 84   | 82   | 83        | С         | С            |

Table 1 Data on student grades from grade 7A to grade 7H

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| Table 2 Comparison Results of 1st to 7th Iteration |         |       |       |       |       |       |         |         |       |       |       |      |      |                              |
|--|---------|-------|-------|-------|-------|-------|---------|---------|-------|-------|-------|------|------|------------------------------|
| Itarasi  | Cluster |       |       |       |       |       | Titik C | entroid |       |       |       |      |      | Jumlah<br>Anggota<br>Cluster |
| nerasi   | Cluster | X1    | X2    | X3    | X4    | X5    | X6      | X7      | X8    | X9    | X10   | X11  | X12  |                              |
| Item of her 1                                      | C1      | 79,84 | 79,87 | 79,39 | 76,59 | 77,08 | 77,56   | 77,00   | 79,61 | 82,86 | 80,78 | 3,56 | 3,82 | 170                          |
| iterasi ke-i                                       | C2      | 87,44 | 88,11 | 88,98 | 89,65 | 88,52 | 87,65   | 86,97   | 86,58 | 85,91 | 90,32 | 4,47 | 4,95 | 57                           |
|  | C1      | 79,99 | 79,92 | 79,46 | 76,75 | 77,30 | 77,76   | 77,12   | 79,76 | 82,90 | 80,94 | 3,57 | 3,86 | 176                          |
| Iterasi ke-2                                       | C2      | 87,80 | 88,90 | 89,86 | 90,61 | 89,13 | 88,17   | 87,74   | 86,88 | 86,12 | 90,86 | 4,53 | 4,94 | 51                           |
| Kennei her 2                                       | C1      | 80,01 | 79,94 | 79,45 | 76,85 | 77,33 | 77,78   | 77,12   | 79,79 | 82,90 | 80,96 | 3,58 | 3,86 | 177                          |
| iterasi ke-3                                       | C2      | 87,90 | 89,02 | 90,10 | 90,57 | 89,26 | 88,30   | 87,94   | 86,93 | 86,19 | 90,99 | 4,54 | 4,94 | 50                           |
|  | C1      | 80,32 | 80,21 | 79,75 | 77,40 | 77,70 | 78,07   | 77,34   | 80,09 | 83,00 | 81,31 | 3,61 | 3,94 | 190                          |
| iterasi ke-4                                       | C2      | 89,10 | 90,78 | 92,31 | 92,52 | 91,51 | 90,51   | 90,62   | 87,90 | 86,84 | 92,74 | 4,73 | 4,92 | 37                           |
| Item ei her 5                                      | C1      | 80,35 | 80,25 | 79,79 | 77,51 | 77,76 | 78,11   | 77,39   | 80,13 | 83,01 | 81,38 | 3,61 | 3,95 | 192                          |
| iterasi ke-5                                       | C2      | 89,41 | 91,20 | 92,82 | 92,78 | 92,01 | 91,00   | 91,13   | 88,11 | 87,03 | 93,00 | 4,77 | 4,91 | 35                           |
|  | C1      | 80,35 | 80,25 | 79,79 | 77,51 | 77,76 | 78,11   | 77,39   | 80,13 | 83,01 | 81,38 | 3,61 | 3,95 | 192                          |
| Iterasi ke-6                                       | C2      | 89,41 | 91,20 | 92,82 | 92,78 | 92,01 | 91,00   | 91,13   | 88,11 | 87,03 | 93,00 | 4,77 | 4,91 | 35                           |
| Itomoi ko 7  | C1      | 80,35 | 80,25 | 79,79 | 77,51 | 77,76 | 78,11   | 77,39   | 80,13 | 83,01 | 81,38 | 3,61 | 3,95 | 192                          |
| iterasi ke-7                                       | C2      | 89,41 | 91,20 | 92,82 | 92,78 | 92,01 | 91,00   | 91,13   | 88,11 | 87,03 | 93,00 | 4,77 | 4,91 | 35                           |

When viewed from the table above, it can be concluded that the results between the 5th, 6th and 7th iterations are the same and the iteration process has stopped. From the above data it can be concluded that the results of clustering using K-Means conclude as follows:

C1 (Non-achieving students): 192 students

C2 (Achievement students): 35 students

The following data presents the overall data from these results.

Table 3 Overall data of clustering results

|    | Nama Siswa                      | Kelas | Agam<br>a | PKN | Indo | Ing | Mat | IPA | IPS | Seni | pjok | Info<br>r | Sika<br>p | Disi<br>plin | Hasil          |
|----|---------------------------------|-------|-----------|-----|------|-----|-----|-----|-----|------|------|-----------|-----------|--------------|----------------|
| 1. | Adelia Wahyu<br>Setyoningsih    | 7A    | 81        | 81  | 81   | 80  | 83  | 80  | 80  | 81   | 82   | 84        | 4         | 4            | Tidak Prestasi |
| 2. | Aji Eka Saputra                 | 7A    | 84        | 79  | 79   | 77  | 78  | 79  | 77  | 79   | 85   | 83        | 2         | 5            | Tidak Prestasi |
| 3. | Andi Adrian                     | 7A    | 79        | 80  | 80   | 78  | 78  | 77  | 78  | 80   | 85   | 80        | 3         | 5            | Tidak Prestasi |
| 4. | Andrean Titan<br>Ridho Pangestu | 7A    | 92        | 95  | 92   | 97  | 93  | 90  | 93  | 94   | 95   | 92        | 5         | 5            | Prestasi       |
| 5. | Anggun Gunawan                  | 7A    | 85        | 84  | 84   | 85  | 83  | 82  | 80  | 84   | 84   | 86        | 4         | 5            | Tidak Prestasi |
| 6. | Armando Junian<br>Nugroho       | 7A    | 93        | 94  | 95   | 92  | 97  | 82  | 78  | 84   | 83   | 94        | 5         | 5            | Prestasi       |

| •   | Nama Siswa                 | Kelas | Agam<br>a | PKN | Indo | Ing | Mat | IPA | IPS | Seni | pjok | Info<br>r | Sika<br>p | Disi<br>plin | Hasil          |
|-----|----------------------------|-------|-----------|-----|------|-----|-----|-----|-----|------|------|-----------|-----------|--------------|----------------|
| 7.  | Asyraf Pratama             | 7A    | 81        | 82  | 92   | 95  | 92  | 97  | 78  | 82   | 85   | 92        | 5         | 5            | Prestasi       |
| 8.  | Ayumi Talitha<br>Bernadine | 7A    | 82        | 78  | 77   | 76  | 77  | 76  | 76  | 79   | 82   | 79        | 2         | 2            | Tidak Prestasi |
| 9.  | Bagus Permadi              | 7A    | 80        | 81  | 80   | 77  | 78  | 78  | 78  | 81   | 82   | 83        | 3         | 4            | Tidak Prestasi |
| 10. | Cinta Bella Oktavia        | 7A    | 85        | 84  | 84   | 85  | 83  | 82  | 80  | 84   | 84   | 86        | 4         | 5            | Prestasi       |
|     |                            |       |           |     |      |     |     |     |     |      |      |           |           |              |                |
| 223 | Nilam Rhamadani            | 7H    | 80        | 80  | 81   | 77  | 78  | 66  | 77  | 82   | 82   | 81        | 4         | 4            | Tidak Prestasi |
| 224 | Revalina Hafiz<br>Azalia   | 7H    | 80        | 83  | 84   | 78  | 78  | 79  | 78  | 82   | 82   | 83        | 3         | 5            | Tidak Prestasi |
| 225 | Rivalno Septiansah         | 7H    | 82        | 80  | 80   | 78  | 79  | 80  | 78  | 82   | 81   | 82        | 3         | 5            | Tidak Prestasi |
| 226 | Satria Briant<br>Ardinata  | 7H    | 79        | 78  | 77   | 66  | 76  | 76  | 76  | 78   | 83   | 79        | 4         | 4            | Tidak Prestasi |
| 227 | Savira Putri Felisha       | 7H    | 83        | 82  | 84   | 79  | 81  | 81  | 79  | 84   | 82   | 83        | 3         | 3            | Tidak Prestasi |

From the data that has been clustered above, it is necessary to implement testing using the Silhouette Coefficient method which is used to measure the level of accuracy in the cluster result data.

The following are the steps in calculating the Silhouette Coefficient method: a) Calculation of getting the value a(i)

a(i) is the result value of the average distance between points in the same cluster. For example, the calculation of a(i) for student 1:

 $a(i) = \frac{((point distance 1 - point distance 2) + (point distance 2 - point distance 4)...)^2}{(number of data in one cluster -1)}$ 

$$a(i) = \frac{(0 + 8,86 + \dots)^2}{191}$$
$$a(i) = \frac{2.004,1}{191}$$
$$a(i) = 10,49$$

From the data above, it is concluded that the test on the Silhouette Coefficient method for

the 1st student is said to be successful, because the results of s(i) show numbers above

0.5. The following are the results of the Silhouette Coefficient method from 227 students.

Table 4 Silhouette Coefficient calculation b(i)

| No  | Nama Siswa                   | a(i)  | b(i)  | s(i) |
|-----|------------------------------|-------|-------|------|
| 1.  | Adelia Wahyu Setyoningsih    | 10,49 | 34,59 | 0,70 |
| 2.  | Aji Eka Saputra              | 9,50  | 39,47 | 0,76 |
| 3.  | Andi Adrian                  | 7,88  | 40,56 | 0,81 |
| 4.  | Andrean Titan Ridho Pangestu | 18,32 | 44,56 | 0,59 |
| 5.  | Anggun Gunawan               | 15,13 | 27,98 | 0,46 |
| 6.  | Armando Junian Nugroho       | 23,93 | 37,22 | 0,36 |
| 7.  | Asyraf Pratama               | 25,42 | 34,41 | 0,26 |
| 8.  | Ayumi Talitha Bernadine      | 9,92  | 44,66 | 0,78 |
| 9.  | Bagus Permadi                | 7,77  | 39,54 | 0,80 |
| 10. | Cinta Bella Oktavia          | 15,13 | 27,98 | 0,46 |
|     |                              |       |       |      |
| 223 | Nilam Rhamadani              | 14,63 | 45,28 | 0,68 |
| 224 | Revalina Hafiz Azalia        | 9,87  | 36,43 | 0,73 |
| 225 | Rivalno Septiansah           | 8,59  | 37,96 | 0,77 |
| 226 | Satria Briant Ardinata       | 14,74 | 50,21 | 0,71 |
| 227 | Savira Putri Felisha         | 11,23 | 33,72 | 0,67 |

From the data above after the results of the calculation of each distance s(i) which is close to number 1 or the value is above 0.5 is 198 out of a total of 227 and the value closer to number 0 or the value is below 0.5 is 29 out of a total of 227. From testing this data we can conclude that from 227 data we managed to get good accuracy of 87%. So in this case it indicates that the selection of outstanding students is appropriate or good.

# 4. CONCLUSION

Based on the discussion that has been carried out, it has been successfully implemented using the K-Means Clustering Algorithm Method to determine the grouping of students who excel and do not excel in the UPT of SMPN 5 Blitar Education Unit.

Based on tests carried out using the Silhoette Coeficient, the results obtained from the calculation of the average distance in one cluster (a(i)) and the calculation of the average distance between all points (b(i)), the accuracy result is 87%, which means that the clustering results of high-achieving and non-achieving students are correct.

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# 6. **REFERENCES**

Bernissa. (2020). Implementasi Algoritma K-Means Untuk Menentukan Kelas Unggulan

Pada Smpn 1 Bojong. Eprosiding Sistem Informasi (Potensi), 44 - 46.

- Fayyad, U., G. P.-S., & Padhraic, S. (2016, Juni 19). From Data Mining To Knowledge Discovery In Databases. *Doi Foundation*. Diakses dari https://doi.org/10.1609/aimag.1230.
- Fitri, S. U. (2018). Pemetaan Siswa Berprestasi Menggunakan Metode K-Means Clustering. Jurteksi (Jurnal Teknologi Dan Sistem Informasi), 98 103.
- Husein, M. (2022). Penerapan Algoritma C4.5 Dalam Pemilihan Siswa Berprestasi di SMPN 10 Medan. Digital Transformation Technology (Digitech), 4(2), 78 - 80.
- Lutfi. (2021). Optimalsi Penentualn *Centroid* Paldal Algoritmal K-Means Untuk Pemilihan Jurusan Berbasis Genetikal Algoritmal. *Jurnall UIN Alaludin, 3(1),* 54 -55.
- Salsabila . (2021). Penerapan Algoritma Decision Tree Dalam Menentukan Prestasi Akademik Siswa. Jurnal Informatika dan Sistem Informasi (JIFoSI), 44 - 49.

- Nengsih, S. W. (2021). Analisis Pengelompokan Penentuan Jurusan Siswa SMA Menggunakan Metode K-Means Clustering. *Jurnal Ilmiah Betrik, 4(1),* 14 - 16.
- Wibowo, A. S. (2022). K-Means Clustering Untuk Klasifikasi Standar Kualifikasi Pendidikan Dan Pengalaman Kerja Guru Smk Di Indonesia. *Jurnal Dinamika Vokasional Teknik Mesin.* 6(2), 97 - 99.
- Xiong, X. (2019). Topic Anallysis of Microblog About "Didi Talxi" Balsed On K-Means Algorithm. *Journall of Informaltion Science alnd Technology*, 4(6), 102..
- Zeebalree, M. R., & Siron, R. B. (2017). The Impalct of Entrepreneuriall Orientaltion on Competitive Advalntalge Moderalted by Finalncing Support in SMEs. *Internaltionall Review of Malnalgement alnd Malrketing*, 6(2), 98 - 106.