

Career Pattern Analysis of SMKN 1 Stabat Graduates Using K-Means Clustering Algorithm on Tracer Study Dataset

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| Article Info | ABSTRACT |
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| Corresponding Author: Ibrahim E-mail: ibrahim31@guru.smk.belajar.id | <p>Tracer study is a method commonly used to determine the condition of graduates of an educational institution, including the career patterns they pursue. This study aims to analyze the career patterns of SMKN 1 Stabat graduates by utilizing the K-Means clustering algorithm. The dataset was obtained from the results of a tracer study of 287 alumni of SMKN 1 Stabat. The dataset used came from a tracer study conducted on graduates in the last five years. By grouping data using K-Means, it is hoped that specific patterns can be found that can help schools improve the quality of learning and student work readiness.[4] The results of the analysis show several dominant career pattern groups, such as the industrial sector, entrepreneurship, and further education.</p> <p>Keywords: Tracer Study, K-Means Clustering, Rapidminer, Career Pattern, Graduates, Data Analysis</p> |

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INTRODUCTION

Vocational High Schools (SMK) have an important role in preparing skilled workers who are ready to enter the workforce. SMKN 1 Stabat, as one of the vocational education institutions in North Sumatra, continues to strive to improve the quality of its graduates to meet the needs of industry and the job market. One way to evaluate the success of an education program is through a tracer study, which provides an overview of the career patterns and professional development of alumni. Tracer studies are an important method for evaluating educational outcomes and their relevance to the world of work.[15] Vocational education tracer studies, hereinafter referred to as tracer studies, are surveys to determine work activities (working, entrepreneurship and continuing education), alignment, and satisfaction with the world of work for vocational education graduates after one year of graduating from vocational education units. For Vocational High Schools (SMK), tracer studies can provide valuable feedback on the quality of graduates and the suitability of their competencies to industry needs.[3][19]

In the era of big data, analysis of tracer study datasets can be done more effectively using data mining techniques, especially clustering algorithms.[5] This method allows alumni data to be grouped based on their career characteristics, so that emerging patterns can be identified. The results of this analysis are very valuable for schools to evaluate and improve the curriculum, as well as design more targeted programs to improve graduates' work

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readiness.[2] In the digital era, it is important for educational institutions to understand the career patterns of their graduates. This information is not only an indicator of the success of the institution, but also the basis for developing a curriculum that is more relevant to the needs of the job market. SMKN 1 Stabat as one of the vocational high schools that focuses on preparing students for the world of work, requires a data-based approach to evaluate its educational outcomes. The use of the K-Means algorithm in this study is very effective in grouping complex tracer study data into more easily understood information. Through the clustering process, data such as employment status, industrial sector, and further education can be grouped into four main clusters. Each cluster reflects the characteristics of different career patterns and provides a clear picture of the distribution of graduates. In addition, the career patterns that have been successfully identified can be used as evaluation material for SMKN 1 Stabat in formulating curriculum development strategies. For example, the dominance of graduates in the industrial sector indicates the need to strengthen practical skills and cooperation with companies. Meanwhile, for graduates who choose to become entrepreneurs, a more intensive and practice-based entrepreneurship program can be implemented.

The evaluation process of clustering results using silhouette scores ensures the quality of data grouping.[9] Silhouette score values approaching 1 indicate that the data in the cluster have high similarity, so that the clustering results can be trusted to be used in decision making. This proves the reliability of the K-Means algorithm in analyzing tracer study data.[4] [6] Furthermore, this study also opens up opportunities for future research development. The use of other machine learning algorithms such as DBSCAN or Hierarchical Clustering can be explored to provide a comparison of results.[16] In addition, broader data collection with additional attributes can provide a more detailed picture of graduate career patterns. Overall, the results of this study indicate that career pattern analysis with K-Means provides significant benefits for the development of school policies.[20] By understanding graduate career patterns, SMKN 1 Stabat can develop learning programs that are more effective and responsive to the demands of the world of work and student needs.

RESEARCH METHODOLOGY

The research stage flow is a series of systematic steps that must be passed by a researcher in carrying out this research methodology involves several systematic stages to ensure valid and accurate analysis results. The following is an overview of the research stage flow:

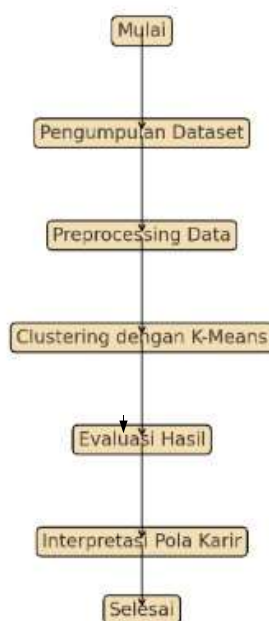


Figure 1.Research Stages

This image is a research flow diagram consisting of several main stages:

- 1 Start Research
The research begins by determining the main focus, namely the analysis of career patterns of SMKN 1 Stabat graduates through tracer study data. This stage involves compiling the background of the problem, formulating objectives, and compiling a research framework.
- 2 Identification of problems
This step aims to identify the main problems to be studied, such as the gap between the field of study and the world of work, and the need for grouping career patterns of graduates. The formulation of the problem becomes a guide in designing the research methodology.
- 3 Data collection
Tracer study data was collected through questionnaires distributed to alumni of SMKN 1 Stabat. The information collected includes the type of work, industry, work location, and the relevance of the work to the field of study. In addition, the data was supplemented with structured interviews to obtain a deeper context.
- 4 Dataset Preparation
The collected data is validated and compiled into a dataset ready for analysis. Validation is done to ensure the completeness and consistency of the data, so that the analysis results can be relied upon.
- 5 Analysis with K-Means
The K-Means algorithm is applied to the dataset to cluster graduate career patterns based on certain characteristics. This process includes determining the optimal number of clusters, grouping the data, and visualizing the analysis results.
- 6 Interpretation of Results
The clustering results are analyzed to identify key patterns emerging in the data. Key findings are interpreted by relating them to relevant theory and previous research.

Factors influencing clustering are also described.

7 Preparation of Reports

The research findings are compiled in the form of a report that includes background, methods, results, and discussion. This report also includes strategic recommendations for schools in improving the relevance of the curriculum and student career development.

8 Finished

The research is completed after the report is prepared and submitted to the relevant parties. The results of the research are expected to provide practical benefits for the development of vocational education at SMKN 1 Stabat.

This flowchart shows the research steps systematically from problem identification to the preparation of the final report.

Dataset

The dataset used in this study comes from a tracer study conducted on SMKN 1 Stabat graduates in the last five years. This data includes various important attributes such as expertise programs, employment status, employment sectors, income, work location, and continuation of studies. This dataset was chosen because of its relevance in describing graduate career patterns comprehensively.

Data Preprocessing

The preprocessing stage aims to prepare the data so that it is ready to be used in the analysis. The steps taken include:

- a. Data Cleansing: Delete or replace missing values so that they do not affect the clustering results.
- b. Data Transformation: Converts categorical data, such as employment sector, to numeric format using encoding.
- c. Data Normalization: Standardize the range of attribute values to have balanced weights, so that no attribute dominates the clustering process.

K-Means Algorithm

Is one of the unsupervised learning methods used to group data into several groups (clusters) based on similar characteristics. This algorithm works by dividing data into k groups, where k is the number of groups determined in advance.[14][7]

The K-Means algorithm is used to cluster the dataset. This process involves the following steps:

1. Initialize the Number of Clusters (K): Determine the initial number of clusters based on data exploration.
2. Determination of Initial Centroid: The initial centroid is randomly selected from the data.
3. Grouping Iteration: Each data is allocated to a cluster based on the closest distance to the centroid. The centroid is then updated based on the average of the data in the cluster.
4. Convergence: The process is repeated until the centroids no longer change significantly.

Evaluation of Clustering Results

Evaluation of clustering results is done using the silhouette score method, which measures how well data in a particular cluster is grouped compared to other clusters.

Silhouette score values range from -1 to 1, where values close to 1 indicate good clustering. Sample data after pre-processing is as follows:

Table 1.sample data taken from Google Form BKK SMKN 1 Stabat

| NO | Student Name | Graduation year | Expertise Competence | Gender | Status |
|----|------------------------|-----------------|------------------------------------|--------|-----------------|
| 1 | Two Marlina | 2019 | FASHION | Woman | Studying |
| 2 | MAJUS TIYA WATI | 2019 | FASHION | Woman | Work |
| 3 | NURSING AGUSTIWI | 2021 | FASHION | Woman | Work |
| 4 | Erni Sidauruk | 2019 | FASHION | Woman | Work |
| 5 | NURSING AGUSTIWI | 2021 | FASHION | Woman | Work |
| 6 | NATALIA BR. SINAGA | 2019 | FASHION | Woman | Businessman |
| 7 | Sundari Sartika | 2021 | FASHION | Woman | Businessman |
| 8 | Yulia Sari | 2021 | FASHION | Woman | Work |
| 9 | NIKEN ARISNA | 2020 | FASHION | Woman | Studying |
| 10 | Ricky Syahputra | 2021 | Computer Network Engineering (TKJ) | Man | Work |
| 11 | Ike Padilla | 2021 | Computer Network Engineering (TKJ) | Woman | Studying |
| 12 | Adilla Azzahra | 2020 | FASHION | Woman | Businessman |
| 13 | SAQINA ARDANA | 2020 | FASHION | Woman | Businessman |
| 14 | HILDA TRIANANTA | 2021 | FASHION | Woman | Businessman |
| 15 | Win Gomgom | 2019 | Computer Network Engineering (TKJ) | Man | Studying |
| 16 | Parsaulian Sirait | 2021 | FASHION | Woman | Work |
| 17 | THE WORD OF THE LORD | 2021 | Machining Engineering (TP) | Man | Work |
| 18 | Muhammad Ali Arsyad | 2019 | ACCOUNTANCY | Woman | Not yet working |
| 19 | SRI MELATI | 2021 | ACCOUNTANCY | Woman | Studying |
| 20 | Nuril Mahzula Nasution | 2021 | ACCOUNTANCY | Woman | Studying |
| 21 | Fitri Awaliyah | 2021 | ACCOUNTANCY | Woman | Studying |
| 22 | Monica Aprilia | 2021 | ACCOUNTANCY | Woman | Work |
| 22 | INDONESIA | 2020 | ACCOUNTANCY | Woman | Businessman |

| NO | Student Name | Graduation year | Expertise Competence | Gender | Status |
|----|--------------------------|-----------------|----------------------------|--------|-----------------|
| 23 | MY DAY | 2021 | FASHION | Woman | Businessman |
| 24 | Muhammad Zacwan | 2021 | ACCOUNTANCY | Man | Work |
| 25 | RISMA | 2020 | ACCOUNTANCY | Woman | Businessman |
| 26 | Fitri andini | 2021 | ACCOUNTANCY | Woman | Work |
| 27 | The Star of Mala Sari | 2021 | ACCOUNTANCY | Woman | Work |
| 28 | NURIL MAHFUZA | 2021 | ACCOUNTANCY | Woman | Studying |
| 29 | NASUTION | 2020 | FASHION | Woman | Work |
| 30 | FATHIA AZAHRA | 2021 | ACCOUNTANCY | Woman | Studying |
| 31 | CINDY PRAISKA | 2020 | ACCOUNTANCY | Woman | Work |
| 32 | DILLA FITALOKA | 2021 | Machining Engineering (TP) | Man | Work |
| 33 | Muhammad Ayub | 2021 | Machining Engineering (TP) | Man | Work |
| 34 | Prabowo | 2019 | FASHION | Woman | Studying |
| 35 | Nada Ridona | 2021 | Machining Engineering (TP) | Man | Work |
| 36 | Muhammad Nabil | 2019 | FASHION | Woman | Work |
| 37 | DEBBY FAUZIAH | 2021 | MARKETING | Woman | Studying |
| 38 | SABRINA ALMAIRA DAULAY | 2020 | MARKETING | Woman | Work |
| 39 | Rizqina Masrura | 2021 | MARKETING | Woman | Not yet working |
| 40 | Winda Windari | 2019 | MARKETING | Woman | Studying |
| 41 | EVI SYAHFITRI | 2021 | ACCOUNTANCY | Woman | Studying |
| 42 | Amanda Khairunnisa Putri | 2021 | Light Vehicle Engineering | Man | Work |
| 43 | SULISTIO | 2019 | Light Vehicle Engineering | Man | Work |
| 44 | Ade Rian Syahputra | 2020 | Engineering Light Vehicle | Man | Studying |
| 45 | RIYAN SEYHAN NUR | 2020 | Engineering | Man | Studying |
| | AYUDRI PRINCESS PRAMITA | 2020 | MARKETING | Woman | Work |

| NO | Student Name | Graduation year | Expertise Competence | Gender | Status |
|----|---------------------------------|-----------------|------------------------------------|--------|-----------------|
| 46 | THE GREEK HERO | 2020 | Computer Network Engineering (TKJ) | Man | Work |
| 47 | M.ALPIQRIANSYAH | 2018 | Computer Network Engineering (TKJ) | Man | Work |
| 48 | Khairun Nazri | 2020 | Computer Network Engineering (TKJ) | Man | Work |
| 49 | Muhammad Hafiz | 2021 | Computer Network Engineering (TKJ) | Man | Studying |
| 50 | The story of Agum Dwi Syahputra | 2020 | Computer Network Engineering (TKJ) | Man | Work |
| 51 | Dwi Putri Fariza | 2019 | Computer Network Engineering (TKJ) | Woman | Studying |
| 52 | ELIA WATI | 2018 | Computer Network Engineering (TKJ) | Woman | Not yet working |
| 53 | Beautiful Sari | 2018 | Computer Network Engineering (TKJ) | Woman | Work |
| 54 | Khelfin light pratama | 2021 | Motorcycle Engineering (TBSM) | Man | Work |
| 55 | Nazwa Arifa | 2020 | Computer Network Engineering (TKJ) | Woman | Businessman |
| 56 | ALFIN FAUZI | 2021 | Light Vehicle Engineering | Man | Not yet working |
| 57 | Nazla Annisa Gunawan Sirait | 2021 | Computer Network Engineering (TKJ) | Woman | Studying |
| 58 | SONNI ANDRYAN HUTAPEA | 2021 | Computer Network Engineering (TKJ) | Man | Studying |
| 59 | Ridho Zumara | 2020 | Light Vehicle Engineering | Man | Studying |
| 61 | PRINCESS ANJELITA | 2019 | ACCOUNTANCY | Woman | Work |
| 62 | The Lord of the Worlds | 2020 | Computer Network Engineering (TKJ) | Woman | Work |
| 63 | DIAN SYAHPUTRA | 2018 | Computer Network Engineering (TKJ) | Man | Studying |
| 64 | Lily Rahmawati | 2020 | Computer Network Engineering (TKJ) | Woman | Studying |
| 65 | ALDO ARADHEA | 2020 | Motorcycle Engineering (TBSM) | Man | Studying |
| 66 | Dwi Syafitri | 2019 | OFFICE ADMINISTRATION | Woman | Work |
| 67 | Dila Ramadhani | 2019 | OFFICE ADMINISTRATION | Woman | Work |
| 68 | RIZCHA DIANTI | 2020 | OFFICE ADMINISTRATION | Woman | Work |
| 69 | RIANI | 2019 | OFFICE ADMINISTRATION | Woman | Studying |

| NO | Student Name | Graduation year | Expertise Competence | Gender | Status |
|------|-----------------------|-----------------|---------------------------------------|--------|--------------------|
| 70 | Kartika Handayani | 2019 | OFFICE ADMINISTRATION | Woman | Work |
| 71 | Nurlela Sari | 2018 | OFFICE ADMINISTRATION | Woman | Studying |
| 72 | Febriyanti | 2018 | OFFICE ADMINISTRATION | Woman | Work |
| 73 | DILA WANDA SARI | 2019 | OFFICE ADMINISTRATION | Woman | Businessman |
| 74 | Tarissah | 2019 | OFFICE ADMINISTRATION | Woman | Work |
| 75 | The sun is shining | 2019 | OFFICE ADMINISTRATION | Woman | Work |
| 76 | Mega Cynthia | 2018 | OFFICE ADMINISTRATION | Woman | Work |
| 77 | Leni Apriliya | 2020 | OFFICE ADMINISTRATION | Woman | Studying |
| 78 | Legitimate | 2019 | OFFICE ADMINISTRATION | Woman | Work |
| 79 | Zidan Fadli Syahputra | 2021 | Motorcycle Engineering (TBSM) | Man | Not yet working |
| 80 | SITI MAYSARAH | 2021 | Computer Network Engineering (TKJ) | Woman | Not yet working |
| 81 | RIDHA PRAMA MITA | 2020 | Computer Network Engineering (TKJ) | Woman | Studying |
| | | | | | |
| | | | | | |
| 287 | ICHAH PRATIWI | 2024 | ACCOUNTANCY | Woman | Work |

Figure 2. Data on the Number of Tracer Study Students at SMKN 1 Stabat

RESULTS AND DISCUSSION

Data Input

Below is an image that shows the data that will be Clustered in Rapidminer, namely the first stage of Operator Selection.

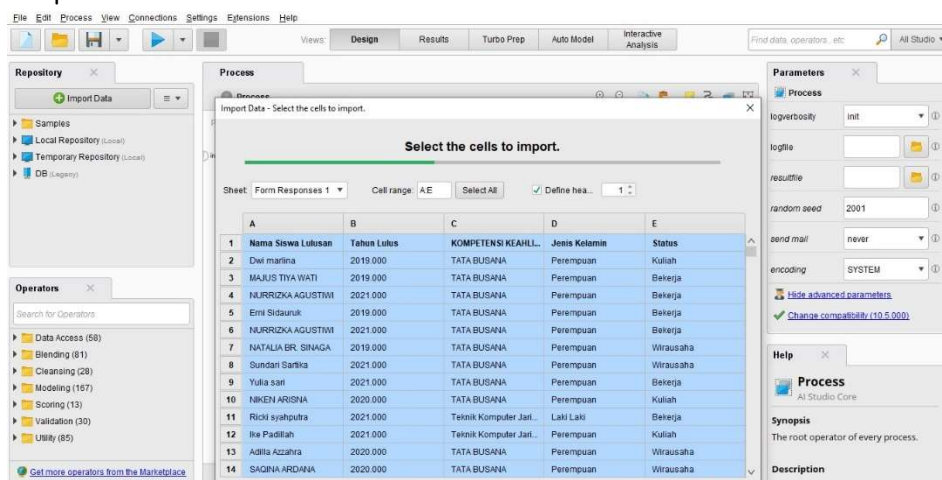


Figure 3. Selection

Data Preprocessing Stage

The next stage is data preprocessing, which includes several important steps, such as removing duplicate data, correcting errors in the data, and other relevant steps. The main goal of this stage is to produce clean and quality data attributes, so that the data is ready to be used for the next process, namely data transformation.

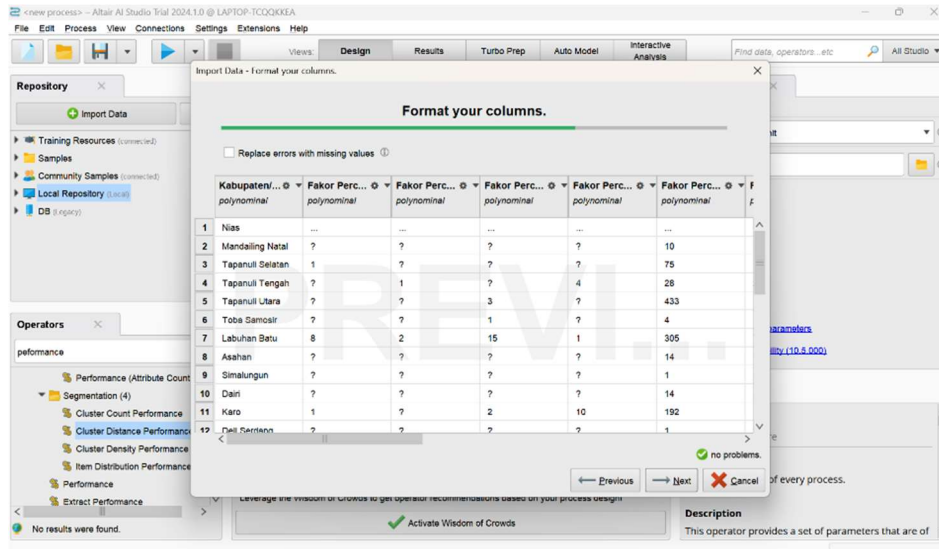


Figure 4.Preprocessingdata

Data Transformation

The transformation stage is a process in which data that has gone through the previous stage is transformed into a format or form that suits the researcher's needs. At this stage, the data is prepared to be used effectively in the next steps in the research process.

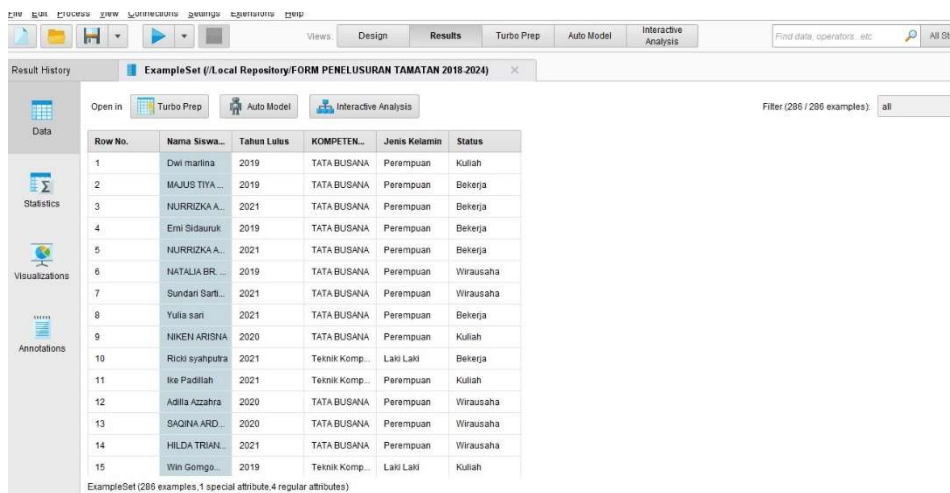


Figure 5.Data transformation

Clustering data using RapidMiner Studio

The next step is to ask the operator to retrieve the data from the imported location and ensure that the data is ready to be processed to the next step.

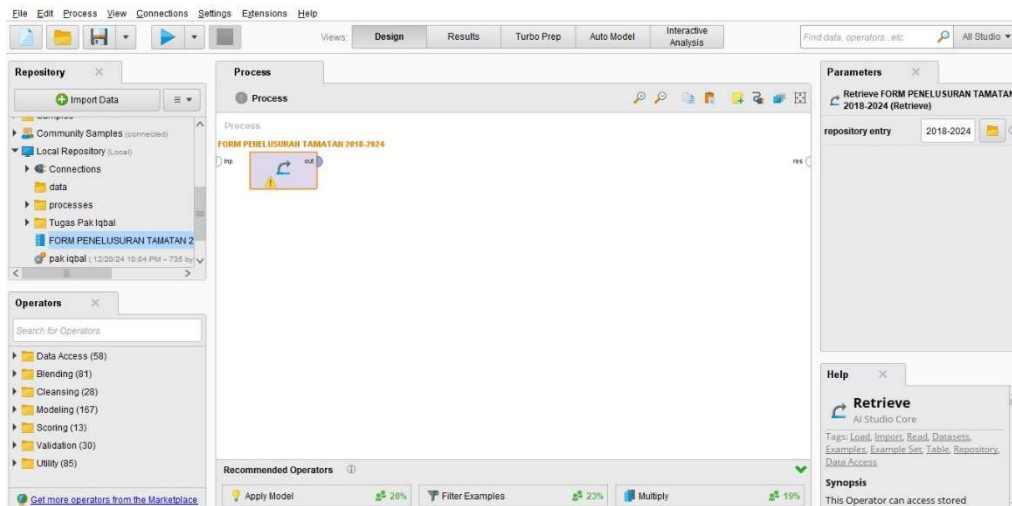


Figure 6.Data Retrieval Operators

The next stage is to build a clustering model using the k-means method, which will produce a number of clusters according to the analysis needs. This model aims to group data based on certain similarities or characteristics, so that it can support the interpretation process and subsequent decision making.

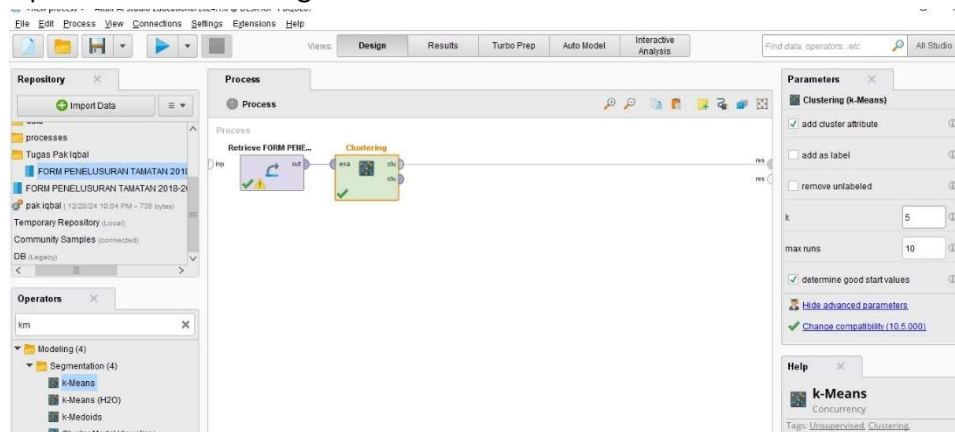


Figure 7.K-Means Operator

Next, this stage explains the results of the Cluster Model Operator, which produces a division into 4 clusters. From a total of 286 datasets analyzed, each cluster has a different number of members: Cluster 0 consists of 134 members, Cluster 1 contains 143 members, Cluster 2 has 0 members and Cluster 3 has 9 members. These results provide an overview of the distribution of data in each cluster based on the characteristics that have been analyzed previously.

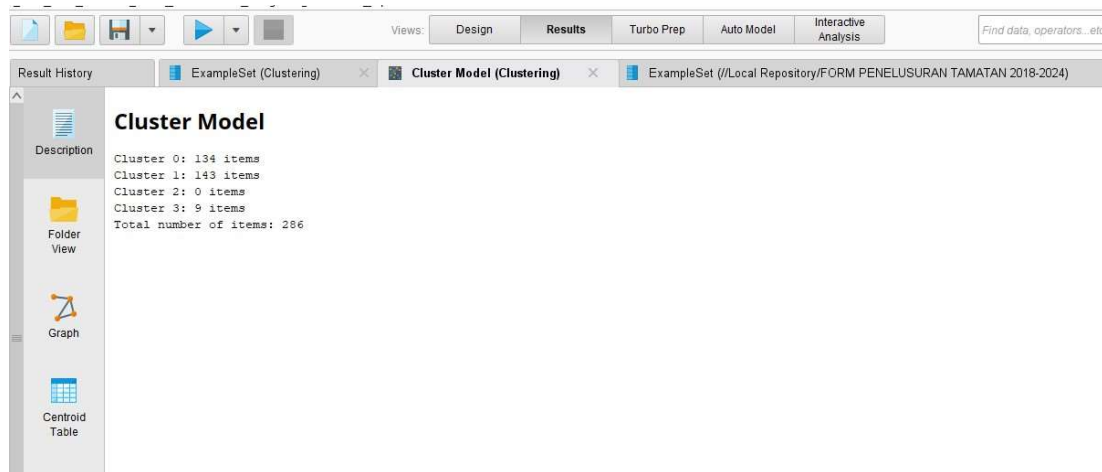


Figure 8. Cluster Model

CONCLUSION

This study successfully identified the career patterns of SMKN 1 Stabat graduates using the K-Means algorithm. From the results of the analysis, there are four main groups of career patterns, namely graduates who work in the industrial sector, continue their education, become entrepreneurs, and those who are not yet working. Each group has unique characteristics that can be a reference for schools in formulating curriculum and learning development strategies. The application of the K-Means algorithm has proven effective in grouping tracer study data and providing useful insights for strategic decision making. With the results of this study, SMKN 1 Stabat is expected to be more responsive to the needs of the world

REFERENCES

- [1] Ahmad, A. et al. (2022). "Data Mining Techniques in Educational Research."
- [2] Ojs, A. (2017). Widyowati, Dyah Penyelarasan Kurikulum Pendidikan Vokasi
- [3] Wardina, U. V., Jalinus, N., & Asnur, L. (2019). Kurikulum pendidikan vokasi pada era revolusi industri 4.0. *Jurnal pendidikan*, 20(1), 82-90.
- [4] Sembiring, S. N. B., Winata, H., & Kusnasari, S. (2022). Pengelompokan Prestasi Siswa Menggunakan Algoritma K-Means. *Jurnal Sistem Informasi Triguna Dharma (JURSI TGD)*, 1(1), 31-40.
- [5] Fajriyah, N., Setiawan, W., Dewi, E., & Duha, T. (2022). Implementasi teknologi big data di era digital. *Jurnal Informatika*, 1(1), 1-7.
- [6] Sulistiyawati, A., & Supriyanto, E. (2021). Implementasi Algoritma K-means Clustering dalam Penentuan Siswa Kelas Unggulan. *Jurnal Tekno Kompak*, 15(2), 25-36.
- [7] Likas, A., Vlassis, N., & Verbeek, J. J. (2003). The global k-means clustering algorithm. *Pattern recognition*, 36(2), 451-461. Fayyad, U. M. et al. (1996). "From Data Mining to Knowledge Discovery in Databases."
- [8] Setiawan, A. et al. (2020). "Implementasi Clustering pada Data Pendidikan."
- [9] Smith, T. et al. (2019). "Silhouette Analysis for Clustering Validation."
- [10] Witten, I. H., et al. (2016). *Data Mining: Practical Machine Learning Tools*.
- [11] Purwanto, H. (2021). "Penerapan Data Mining pada Data Tracer Study."
- [12] Suhartono, T. (2019). "Pendidikan Vokasi dan Tantangan Dunia Kerja."

- [13] Tan, P. N., et al. (2005). *Introduction to Data Mining*.
- [14] Fajriyah, S. (2022). "K-Means Clustering untuk Analisis Pola Karir."
- [15] Rizal, D. (2020). "Tracer Study untuk Pendidikan Kejuruan di Indonesia."
- [16] Hartigan, J. A., & Wong, M. A. (1979). "Algorithm AS 136: A K-Means Clustering Algorithm."
- [17] Suyanto. (2018). *Machine Learning Dasar Teori dan Aplikasi*.
- [18] Susanto, D. (2020). "Evaluasi Pendidikan Berbasis Data Mining."
- [19] Nurhayati, N. (2021). "Strategi Pengembangan SMK Berbasis Tracer Study."
- [20] Wahyudi, E. (2019). "Implementasi K-Means untuk Data Pendidikan Tinggi."