# GROUPING OF SEWING TOOL ASSISTANCE RECIPIENTS USING K-MEANS CLUSTERING ANALYSIS

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## **Article Info**

# ABSTRACT

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sewing equipment assistance classification distribution k-means method Aid programs for underprivileged communities need continuous data collection in overcoming welfare problems, what has happened so far is by providing direct assistance to very poor families in every village in Indonesia. Related to this, the problem that has occurred so far is that the direct assistance program is not right on target, because many deserving families should not receive the assistance. The data obtained from the village government is not accurate, it is found that data is considered invalid. This study aimed to determine the distribution of sewing equipment recipients in the best cluster, for sewing equipment recipients in Greged Village, Cirebon Regency. As one way to improve data accuracy, a computational method or model is needed in the form of a data mining algorithm using the k-means clustering method to generate priority groups among hundreds of citizens or the poor. The stages start from data collection, training data, and testing data that consider several criteria from household information, economic conditions, housing conditions, and the number of household members in Greged Village, Cirebon Regency. The results of the tests carried out using 155 data with the best level of accuracy were in the K3 cluster with the Davies Bouldin Index's value o: - 0.584. With the K-Means method, it is very appropriate to determine the recipient of the sewing equipment program in Greged Village.

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#### 1. INTRODUCTION

Determination of recipients of sewing equipment assistance for the community still uses conventional methods to collect data on residents or potential beneficiaries of sewing equipment. The collected data is then carried out in the selection stage. The selection carried out by the Village Government is also still using the manual method and is still not integrated, namely by comparing and separating data from residents or prospective recipients who meet the criteria and those who do not meet the criteria. (Hendarman Lubisa, Ratna Salkiawatib, Sudirman Hala, 2022).

The problem that has occurred so far is that the target in the distribution of the sewing tool assistance program is not accurate, many eligible families should not receive the assistance due to inaccurate data obtained from village government officials found data that is considered invalid. As one way to improve data accuracy, a computational model or method is needed in the form of a data mining algorithm using the K-Means Clustering method. Poverty is a multidimensional problem characterized by the population's low average quality of life, education, health, children's nutrition, and drinking water sources. (Mayssara A. Abo Hassanin Supervised, 2014).

The purpose of this study was to determine the distribution of sewing equipment assistance program groups for the community through social assistance. In predicting the acceptance of sewing equipment assistance using predetermined criteria, to determine whether recipients of assistance were eligible or not eligible to receive community assistance. K-Means Method Clustering has classification capabilities such as decision tree and neural network methods. This method can be used to predict the probability of membership in a class (Sugianto & Maulana, 2019).

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The research conducted by Dede Witan, et al, entitled Classification of Determination of Beneficiaries of the Expected Family Program (PKH) Using the C5.0 Algorithm concluded that the C5.0 Algorithm can be used as a classification method in supporting the decision of beneficiaries of the Expected Family Program by taking into account the gain value. The highest (reinforcement) of the four attributes is ownership of school children, type of house walls, ownership of toddlers, and types of work usually done. The decision tree generated by the classification technique Algorithm C5.0 from the variable that has the highest gain, namely having school children can be used as a reference in determining the beneficiaries of the family program of hope in the future.

From the process of applying data mining using the K-Means method in determining the recipient of sewing equipment, using RapidMiner Software, because it has a high level of accuracy and provides predictions of the level of community welfare based on data updating an integrated database. This study resulted in the best level of accuracy in the K-3 cluster with the Davies Bouldin Index of -.584%.

#### 2. LITERATURE REVIEW

#### 2.1. Data Mining

Data mining is a term used to describe the process of extracting values from the database and is one solution to explain the process of extracting information in a large-scale database and the process of case automation classification based on data patterns obtained from datasets. Data mining is also defined as a set of techniques that are used automatically to thoroughly explore and bring to the surface complex relationships in very large data sets. (Apolinaria Ifon Purnama1), Abdul Aziz2), Anggri Sartika Wiguna3), 2019).

#### 2.2. K-Means method

The k-means method is a method that is included in the distance-based clustering algorithm that divides data into several clusters and this algorithm only works on Numeric attributes. (Windha Mega Pradnya Dhuhita, 2015).

#### 2.3. Clustering

Cluster analysis is finding a collection of objects so that objects in one group are the same or have a relationship with others and are different or not related to objects in other groups. (Moh. Fadel Asikin1, Dian Eka Ratnawati2, Mochammad Ali Fauzi3, 2018).

#### **RESEARCH METHOD** 3.

K-Means is the most famous clustering method and is widely used in various fields because it is simple, easy to implement, can cluster large data, and has linear time complexity where n is the number of documents, K is the number of clusters, and T is the number of clusters. iteration. K-means is a partitioning method of clustering that separates data into different groups. By iterative partitioning, K-Means can minimize the average distance of each data to its cluster. (1M. Nanda Variestha Waworuntu, 2Muhammad Faisal Amin, 2018)

The basic K-means algorithm is as follows:

$$v = \sqrt{\frac{\sum_{i=1}^{n} Xi}{n}}$$

Where: i: 1, 2, 3,... n v: centroid on cluster xi: object i n: the number of objects/number of objects that are members of the cluster Calculate the distance of each data to each centroid using the formula.

$$d(x,y) = \sqrt{\sum_{i=1}^{n} (x_i - y_i)^2}$$

-Group each data based on the shortest distance between the data and its centroid (C). -Return to step 3 if the position of the new centroid with the old centroid is not the same. Information:

dij: object distance between x and y objects

.....

n: Number of Attributes x<sub>i</sub>: Data Object yi: Data Cluster

#### **RESULTS AND ANALYSIS** 4.

# **Data Collection Results**

The purpose of the grouping carried out in this study is how to group the data on sewing equipment recipients for their dataset sourced from the Central Statistics Agency (BPS) with analysis using the K-Means method.

This study uses raw data from the Greged Village Government in 2017 which has 7 attributes and 155 records. The researcher's sample dataset presents 20 records from 155 records as shown in Table 4.1

No	Name	Dusun	Rt	Rw	Desa	Districts
1	Elah	Dusun Pahing	1	3	Greged	Greged
2	Erlianawati	Dusun Pahing	1	3	Greged	Greged
3	Iin	Dusun Pahing	2	3	Greged	Greged
4	Asinah	Dusun Pahing	2	3	Greged	Greged
5	Ucih	Dusun Pahing	3	3	Greged	Greged
6	Robiah	Dusun Pahing	4	3	Greged	Greged
7	Kuniah	Dusun Pahing	4	3	Greged	Greged
8	Yati Maryati	Dusun Pahing	2	3	Greged	Greged
9	Wasniah	Dusun Pahing	1	3	Greged	Greged
10	Ela Jamila	Dusun Pahing	1	3	Greged	Greged
11	Rina Rinawati	Dusun Pahing	2	3	Greged	Greged
12	Ajeung	Dusun Pahing	2	3	Greged	Greged
13	Anisah	Dusun Pahing	3	3	Greged	Greged
14	Feti	Dusun Pahing	4	3	Greged	Greged
15	Neti	Dusun Pahing	2	3	Greged	Greged
16	Nia	Dusun Pahing	2	3	Greged	Greged
18	Wawat	Dusun Pakuwon	1	2	Nanggela	Greged
19	Iis	Dusun Dukuhsari	2	2	Nanggela	Greged
20	Endang Susanti	Dusun Cikaramat	1	2	Nanggela	Greged

Table 4.1. Raw data samples

#### **Data Mining**

In this data mining process, the k-means algorithm is used to classify sewing equipment assistance in Greged Village, Cirebon Regency. In this research, data processing of sewing equipment uses RapidMiner software using the k-means algorithm, the stages is as follows:

1. Open RapidMiner, then it looks like in Figure 4.1:

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<ul> <li>Uturby (85)</li> </ul>	RapidMiner Studie	Core						
Extensions (17)	Symposis							

#### Figure 4.1 RapidMiner Main Screen

Figure 4.1 shows the main page of RapidMiner, which is the display for creating a work page on RapidMiner. Meanwhile, data import is done to enter data to be tested in .xls or .xlsx format. The steps for importing data are:

- Right-click on the main screen of RapidMiner .
- Insert Operator
- Data Access •
- Files •
- Read
- Read Excel, it looks like Figure 4.2: .

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Figure 4. 2 Calling the Read Excel Operator

# **Read Excel Operators**

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Blending (77)		🖌 Activate	Wisdom of Crow	ds		
Cleansing (26)						
Modeling (129)	Help ×					
Validation (29)						
Utility (85)	Read Excel					
Educations (dT)	KapioMiner Studio Con	e				
EXHIBIORS (17)						

Figure 4.3 Read Excel operators that have been called

The next stage after the read excel operator has been called, the next step is to import data for processing by:

- click Import Configurations Wizard
- click the file you want to use. If the data has been successfully imported and there are no further errors,

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Crossref

#### • click finish.

The following is a display of import data as shown in Figure 4.4 and Figure 4.5



Figure 4. 4 Data import wizard

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Repository ×	7	This wizar Step 4: Ra Furthermo operators	d guides pidMinei re, Rapii These r	you to import yo Studio uses st Miner Studio a bles can be also	our data rongly tj ssigns i o define	ped attribute oles to the a d here. Final	is. In this st tributes, de ly, you can r	ep, you o fining wi rename a	an defin hat they attributes	e the data can be us s or deselv	types of your ed for by the i ect them enti	rattributes. ndividual ely.	
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Figure 4. 5 Data import wizard

#### Nominal to numerical operators

The nominal to numerical data operator is to carry out simple validation by dividing the dataset randomly into two separate data, namely training data and test data along with the split data operator display as shown in Figure 4.6:

🔰 📒 - 🔁	C <sup>+</sup> Views: Design Results		() Question	57 <b>v</b>
Repository ×	Process ×	Parameters	×	
😋 Add Data 🛛 = 👻	🕘 Process 100% 🔑 🔑 📮 🍹 💣 🔛	Process		
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Numerical to Binominal		Help ×		
Numerical to Polynomin		Process	5	
Nominal to Binominal		Synopsis		
Nominal to Text	Laurenze Ber Mitchen af Present fa ant anneter anneter antifaine ber af an ann anness derived	The root operat	tor which is the	
Nominal to Numerical	Leverage me visuum or unomus to get operator recommendadors based on your process design?	outer most ope	rator of every	
	Activate Wisdom of Crowds	process.		

Figure 4.6 Nominal to the numerical operator

#### **Clustering operators**

The clustering operator is to specify or set the k:1 to cluster k= to the nth cluster. is to determine the accuracy of the Davies Boulding Index (dbi) data as shown in Figure 4.7

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Figure 4. 7. Clustering operators

### **Cluster distance performance**

Cluster distance performance is this operator reads the Example Set from the specified excel file. Display cluster operator distance performance is presented as shown in Figure 4.8



# **Cluster model**

The k-means process is the process stage for testing the read excel dataset which consists of the training data set and the testing data set connected to the read excel operator. The data model cluster is displayed as shown in Figure 4.9 below:

	ExampleSet (Nominal to Numerical) 🛛 🛛 🗙
	Result History ×
	Cluster Model
Description	Cluster 0: 18 items
	Cluster 1: 15 items
	Cluster 2: 12 items
	Cluster 3: 17 items
Folder	Cluster 4: 14 items
View	Cluster 5: 17 items
	Cluster 6: 14 items
	Cluster 7: 15 items
	Cluster 8: 17 items
Croph	Cluster 9: 16 items
Graph	Total number of items: 155

Figure 4.9 Process Cluster model

The explanation of the cluster model from Figure 4.9 which produces a sequence of clusters includes cluster 0: totaling 18 items, cluster 1: totaling 15 items, cluster 2: totaling 12 items, cluster 3: totaling 17 items, cluster 4: totaling 14 items, cluster 5: totaling 17 items, cluster 6: totaling 14 items, cluster 7: totaling 15 items, cluster 8: totaling 17 items, cluster 9: totaling 14 items and cluster 9: totaling 14 clusters.

#### Data example set

The data example set is used to apply a previously trained model using training data on unlabeled data (data testing). The goal is to get predictions on unlabeled data (testing data) that do not yet have a label. What needs to be considered is that the testing data must have the same order, type, and attribute roles as the training data. The following is a sample data set table display as shown in Figure 4.10.

	-	$\ominus$ $\ominus$	••	Views:	Design	Results			
	Result History	×			📓 Cluster	Model (Clustering	0 ×		
	ExampleSet	Nominal to Nume	rical) $\times$		🐒 Perl	ormanceVector (F	Performance)		
-	ExampleSet (15	55 examples, 2 spec	ial attributes, 56 re		Filter (155 / 1	155 examples):	all	٠	
Data	Row No.	id	cluster	DUSUN - DU	DUSUN - DU	DUSUN - DU	DU SUN - DU	DUSUN - DU	1
Cala	1	1	cluster_0	1	0	0	0	0	ď
Σ	2	2	cluster_0	1	0	0	0	0	¢
	3	3	cluster_0	1	0	0	0	0	¢
Statistics	4	4	cluster_0	1	0	0	0	0	¢
	5	5	cluster_0	0	1	0	0	0	¢
	6	6	cluster_0	1	0	0	0	0	¢
Charts	7	7	cluster_0	1	0	0	0	0	¢
	8	8	cluster_0	1	0	0	0	0	¢
-	9	9	cluster_0	1	0	0	0	0	¢
Advanced	10	10	cluster_0	1	0	0	0	0	¢
Charts	11	11	cluster_0	1	0	0	0	0	(
	12	12	cluster_0	1	0	0	0	0	¢
	13	13	cluster_0	1	0	0	0	0	¢
Annotations									Y

Figure 4.10 Data example set

### **Display Average centroid distance**

The display of the average centroid distance as shown in Figure 4.11



Figure 4. 11 Data example set

#### **Cluster Model Davies Bouldin Index**

Display of the Davies Bouldin Index Cluster Model as shown in Figure 4.12

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Figure 4.12 Results of Davies Bouldin index

# Display cluster graph

Cluster graph display as shown in Figure 4.13



Figure 4.13 Data example set cluster

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#### **Davies Bouldin Index (dbi) process**

k	Dbi	
2	0.587	
3	0.584	
4	0.738	
5	0.713	
6	0.737	
7	0.724	
8	0.851	
9	0.887	
10	0.644	
11	0.787	

Figure 4.14 Davies Bouldin Index Process

# Explanation: Figure 4.14

After carrying out the Cluster distance performance process, the resulting K-means clustering k=10, can produce davies bouldin with sequence k2=0,587, sequence k3=0,584, sequence k4=0,588, sequence k5=0,581, sequence k6=0,529, sequence k7=0,612, the order of k8 = 0.653, the order of k9 = 0.639, the order of k10 = 0.669, then the conclusion of the Cluster distance performance process produces the best dbi of 0.584 in the 3rd order.

#### CONCLUSION

From the process of implementing data mining through the cluster distance performance process, the resulting K-means clustering k=10, can produce a davies Bouldin index with the order k2=0.587, the order k3=0.584, the order k4=0.588, the order k5=0.581, the order k6=0.529, sequence k7 = 0.612, sequence k8 = 0.653, sequence k9 = 0.639, sequence k10 = 0.669, then the conclusion of the Cluster distance performance process produces the best dbi of 0.584 in the 3rd order.

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