

Implementation of Data Mining on Rice Imports by Major Country of Origin Using Algorithm Using K-Means Clustering Method

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ABSTRACT

Indonesia is a country where most of its people rely on the agricultural sector as a livelihood. Indonesia's rice production is so high that it cannot meet the needs of its population. Consequently Indonesia still has to import rice from other food producing countries. One of the leading causes is the enormous population. Statistics show that in the range of 230-237 million people, the staple food of all residents is rice, so it is clear that the need for rice becomes very large. This study discusses the application of data mining on rice import by leading country of origin using K-Means Clustering Method. Sources of data of this study were collected based on import declaration documents produced by the Directorate General of Customs and Excise. Also since 2015, import data also comes from PT. Pos Indonesia, records of other agencies at the border, and the results of cross-border maritime trade surveys. The data used in this study is the data of rice imports by country of origin from 2000-2015 consisting of 10 countries namely Vietnam, Thailand, China, India, Pakistan, United States, Taiwan, Singapore, Myanmar, and Others. Variable used (1) total import of rice (net) and (2) import purchase value (CIF). The data will be processed by clustering rice imports by leading country of origin in 3 clusters, ie high imported cluster, medium introduced cluster, and low import level cluster. The clustering method used in this research is the K-Means method. Centroid data for high import level clusters 7429180 and 2735452,25, Centroid data for medium import level clusters 1046359.5 and 337703.05 and Centroid data for low import level clusters 185559.425 and 53089.225. The result is an assessment based on rice import index with two high imported cluster countries namely Vietnam and Thailand, four medium-level groups of moderate import countries namely China, India, Pakistan and Lainnya and four low imported cluster countries namely USA, Taiwan, Singapore, and Myanmar. The results of the research can be used to determine the amount of rice imported by the leading country of origin.

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I. Introduction

Indonesia is a country where most of its people rely on the agricultural sector as a livelihood. Indonesia's rice production is so high that it cannot meet the needs of its population. Consequently, Indonesia still has to import rice from other food producing countries. One of the leading causes is the enormous population. Statistics show that in the range of 230-237 million people, the staple food of all residents is rice, so it is clear that the need for rice becomes very large[1].

Cluster analysis is a multivariant technique with the primary objective of grouping objects based on the characteristics they possess. Today, cluster analysis has been applied in many fields written in various studies and journals[2]. In the clustering method, the central concept is emphasized the iterative cluster center search, where the center of the cluster is determined by the minimum distance

of each data in the center of the group [3]. The data used in this study are based on import importation information documents produced by the Directorate General of Customs and Excise through the website <https://www.bps.go.id>. In this case, the researcher raised the topic of rice import by leading country of origin where the process of the method is clustering[4]. The results of the cluster can be used as input for the Indonesian state as a form of mapping of the leading country of origin. The mapping process can be clustered into 3 (three) clusters, i.e., the highest import production, medium import production and low import production[5].

Endang Sugiharti in his research conducted clustering of lecturers data related to the activities and their performance by the implementation and responsibility of the K-Means method[6]. Cluster in the research into the cluster Networking, Software Engineering, and E-Learning. The clustering method used in this research is the K-Means method[7].

K-means can also be defined as a Clustering method that is included in the partitioning approach. The K-Means algorithm is a centroid model[8]. Centroid mode is a model that uses centroid to create clusters. The centroid is the midpoint of a cluster. The centroid is a value. The centroid is used to calculate the distance of a data object against the centroid[9].

A data object is included in the cluster if it has the shortest distance to the cluster's centroid. K-Means algorithm can be interpreted as a simple learning algorithm to solve a grouping problem that aims to minimize double faults [10]. The purpose of this research is to apply K-Means in clustering rice imports by leading country of origin[11].

II. Method

In this study, the technique used is the method of data mining as follows. (A) Data collection stage, (b) Data processing stage, (c) Clustering stage and (d) Stage Analysis. Scenes in the method are further described as follows: In the application of data mining of imported rice production by the first country of origin, relevant data is required. Sources of research data obtained from the data collected by documents of importation of imports produced by the Directorate General of Customs and Excise through the site <https://www.bps.go.id>. The data used in this study is the data of rice imports by country of origin from 2000-2015 consisting of 10 countries namely Vietnam, Thailand, China, India, Pakistan, United States, Taiwan, Singapore, Myanmar, and Others. Variable used (1) the amount of rice import (net) and (2) the value of import purchases (CIF). The data will be processed by clustering rice imports based on central destination countries in 3 clusters, i.e., high imported cluster, medium introduced cluster, and low imported cluster level. The data that has been processed will be processed first to be clustered. In the previous stage, the data of each country of origin of rice imports will be summed in every aspect so that at this stage has been obtained the calculation of the value to be processed at the clustering stage. Clustering is an unattended classification and is a process of partitioning a set of data objects from one set into multiple classes. This can be done by applying various equations and steps about the distance of the algorithm, i.e., with Euclidean Distance. Cluster analysis is a method used to divide the data set into groups based on predetermined similarities. In determining the cluster based on the data already available, it takes a flowchart to facilitate in determining the flow of calculation as a groove to find the results of the application of the cluster to the data to be processed. Here is a flowchart in determining the cluster with K-Means. At this stage, the data analysis of rice imports by country of origin with the primary application of the tool. Rapid Minner. Rapid Miner is a machine learning environment for mining, text mining and predictive analytics. The data obtained is processed by using the weight calculation of each index. In the preceding stages, it has been determined to be clustered into 3 clusters of high imported clusters; medium introduced clusters and low import level clusters. At this stage will be analyzed the results.

III. Result and Discussion

In clustering, the data obtained will be calculated in advance based on the number of rice imports in 2000-2015 based on the leading country of origin. The sum result based on two assessment criteria is net weight and CIF value as shown in table 1.

Table 1. Rice Import Data by Main Country of Origin, 2000-2015

| Country of Origin | 2000 | 2001 | 2002 | 2003 | 2004 | 2005 | 2006 | 2007 | 2008 | 2009 | 2010 | 2011 | 2012 | 2013 | 2014 | 2015 |
|----------------------------|--------------------|------------------|--------------------|--------------------|------------------|------------------|------------------|--------------------|------------------|------------------|------------------|--------------------|--------------------|------------------|------------------|------------------|
| Netto : Ton | | | | | | | | | | | | | | | | |
| Vietnam | 369 546,8 | 142 511,8 | 561 728,9 | 506 012,8 | 58 810,1 | 44 772,5 | 272 832,7 | 1 022 834,6 | 125 070,5 | 20 970,5 | 467 369,6 | 1 778 480,6 | 1 084 782,8 | 171 286,6 | 306 418,1 | 509 374,2 |
| Thailand | 361 734,8 | 189 655,9 | 418 697,6 | 492 114,2 | 129 421,5 | 126 408,9 | 157 983,3 | 363 640,1 | 157 007,3 | 221 372,6 | 209 127,8 | 938 695,7 | 315 352,7 | 94 633,9 | 366 203,5 | 126 745,7 |
| Tiongkok ¹ | 476 776,7 | 24 728,5 | 126 768,4 | 54 440,4 | 110,5 | 1,3 | 100,0 | 901,4 | 3 341,7 | 5 167,6 | 3 637,4 | 4 674,8 | 3 099,3 | 639,8 | 1 416,7 | 479,9 |
| India | 0,1 | 2 047,1 | 405 032,2 | 108 797,4 | 923,3 | 327,0 | 720,6 | 3 571,8 | 289,5 | 473,1 | 601,3 | 4 064,6 | 259 022,6 | 107 538,0 | 90 653,8 | 34 167,5 |
| Pakistan | 20 139,1 | 26 110,0 | 32 281,4 | 49 071,0 | 0,0 | 0,0 | 904,3 | 4 603,6 | 751,3 | 501,5 | 4 992,1 | 14 342,3 | 133 078,0 | 75 813,0 | 61 715,0 | 180 099,5 |
| Amerika Serikat | 49 405,2 | 177 889,1 | 13 392,9 | 107 607,6 | 16 766,7 | 2 184,2 | 801,0 | 821,7 | 1 411,2 | 1 323,4 | 1 644,1 | 2 074,1 | 2 445,5 | 2 790,4 | 1 078,6 | 0,0 |
| Taiwan | 0,0 | 0,0 | 3 541,5 | 9 600,5 | 10 600,0 | 0,0 | 2 500,0 | 625,0 | 0,0 | 0,0 | 0,0 | 5 000,0 | 0,0 | 1 240,0 | 840,0 | 0,0 |
| Singapura | 2 496,9 | 7 268,7 | 21 809,7 | 4 314,8 | 6 614,8 | 6 632,1 | 1 568,4 | 468,3 | 898,2 | 250,0 | 10,8 | 1 506,5 | 22,5 | 0,5 | 0,0 | 0,0 |
| Myanmar | 198,4 | 25 441,3 | 111 687,3 | 41 398,9 | 2 500,0 | 0,0 | 0,0 | 0,0 | 0,0 | 0,0 | 0,0 | 1 140,0 | 11 819,6 | 18 450,0 | 15 616,0 | 8 775,0 |
| others | 75 367,9 | 49 080,6 | 110 439,9 | 55 148,2 | 11 119,8 | 9 250,6 | 698,3 | 9 381,2 | 919,8 | 414,5 | 198,4 | 497,8 | 749,3 | 272,5 | 222,0 | 1 959,2 |
| amount | 1 355 665,9 | 644 732,8 | 1 805 379,9 | 1 428 505,7 | 236 866,7 | 189 616,6 | 438 108,5 | 1 406 847,6 | 289 689,4 | 250 473,1 | 687 581,5 | 2 750 476,2 | 1 810 372,3 | 472 664,7 | 844 163,7 | 861 601,0 |
| Nilai CIF: 000 US\$ | | | | | | | | | | | | | | | | |
| Vietnam | 77 075,6 | 23 440,6 | 112 965,1 | 96 273,1 | 13 544,1 | 12 387,4 | 83 773,4 | 335 558,8 | 47 392,1 | 7 936,9 | 232 915,7 | 946 490,1 | 564 925,7 | 97 303,3 | 143 536,0 | 202 563,1 |
| Thailand | 83 408,0 | 33 756,6 | 82 566,1 | 110 447,0 | 33 023,2 | 32 489,0 | 45 399,3 | 122 425,9 | 64 721,0 | 81 959,8 | 109 133,7 | 533 001,9 | 186 171,4 | 61 787,5 | 175 387,4 | 66 772,4 |
| Tiongkok ¹ | 107 428,4 | 4 645,6 | 25 623,5 | 11 131,6 | 42,5 | 4,0 | 145,0 | 1 500,6 | 6 642,4 | 13 697,0 | 12 728,5 | 15 467,1 | 11 205,6 | 1 526,5 | 4 101,5 | 1 631,0 |
| India | 0,1 | 303,7 | 61 148,0 | 17 282,7 | 423,1 | 317,7 | 644,2 | 1 606,2 | 337,6 | 791,5 | 1 767,5 | 6 307,9 | 122 189,0 | 44 989,1 | 34 299,5 | 13 671,7 |
| Pakistan | 5 231,1 | 5 045,8 | 5 464,8 | 8 198,1 | 0,0 | 0,0 | 188,2 | 1 247,0 | 259,2 | 160,7 | 1 765,8 | 6 053,4 | 52 483,4 | 29 996,9 | 23 909,3 | 62 949,2 |
| Amerika Serikat | 13 828,9 | 52 888,6 | 3 579,3 | 22 830,5 | 5 459,6 | 778,7 | 646,9 | 791,1 | 1 796,3 | 2 005,7 | 1 745,5 | 2 489,6 | 2 718,6 | 2 983,6 | 1 294,3 | 0,0 |
| Taiwan | 0,0 | 0,0 | 521,6 | 3 358,1 | 3 786,5 | 0,0 | 935,6 | 223,6 | 0,0 | 0,0 | 0,0 | 1 050,0 | 0,0 | 465,6 | 252,0 | 0,0 |
| Singapura | 484,3 | 1 405,4 | 3 672,8 | 821,4 | 1 698,4 | 1 880,3 | 516,2 | 170,9 | 465,5 | 100,0 | 27,6 | 981,9 | 32,2 | 1,4 | 0,0 | 0,0 |
| Myanmar | 13,8 | 3 822,2 | 16 969,3 | 6 502,0 | 427,1 | 0,0 | 0,0 | 0,0 | 0,0 | 0,0 | 0,0 | 419,2 | 4 754,5 | 6 498,9 | 5 082,8 | 2 732,3 |
| others | 31 659,8 | 9 604,2 | 30 016,5 | 14 578,3 | 3 348,3 | 3 641,9 | 371,7 | 4 195,3 | 2 528,8 | 1 501,6 | 700,8 | 902,3 | 1 142,8 | 449,2 | 315,7 | 1 282,5 |
| amount | 319 130,0 | 134 912,8 | 342 527,1 | 291 422,9 | 61 752,8 | 51 499,0 | 132 620,5 | 467 719,4 | 124 142,8 | 108 153,3 | 360 785,0 | 1 513 163,5 | 945 623,2 | 246 002,1 | 388 178,5 | 351 602,2 |

The data is then accumulated based on two criteria, i.e., net weight and (2) CIF value as shown in table 2 below:

Table 2. Rice Import Accumulation Data

| No | Country of Origin | Berat Netto(Ton) | Nilai CIF (US\$) |
|----|-------------------|------------------|------------------|
| 1 | Vietnam | 8973929,9 | 3336257,3 |
| 2 | Thailand | 5884429,9 | 2134647,2 |
| 3 | Tiongkok1 | 1582007,8 | 463422,3 |
| 4 | India | 1019299,9 | 305743,5 |
| 5 | Pakistan | 732306,3 | 234729 |
| 6 | Amerika Serikat | 455790,5 | 154904,5 |
| 7 | Taiwan | 70968,5 | 20237,8 |
| 8 | Singapura | 3156,8 | 1779,5 |
| 9 | Myanmar | 212321,9 | 35435,1 |
| 10 | Lainnya | 851824 | 346917,4 |

Once accumulated it will get the value of all rice imports by the destination country. Then the data will go into the clustering stage by applying the K-Means algorithm using rapid manner to cluster the data into three clusters. Accumulated data will be entered into the Rapid Minner tool. So it can be clustered into two. Once introduced into Rapid Minner. The results of data accumulation can be seen in table 2. In the application of the K-means algorithm the value of midpoint or centroid is obtained from the data collected with the provision that the desired clusterization is 3, the cluster determination is divided into three parts namely the high import level cluster (C1), the medium import cluster (C2) and the cluster level Low import (C3). Then the value of the midpoint or centroid also there is 3 points. The determination of the cluster point is carried out by taking the largest (maximum) amount for the high imported level cluster (C1), the average cost for the medium imported cluster (C2) and the smallest (minimum) amount for the low import level cluster (C3). The point value can be seen in Table 3 below:

Table 3. Centroid Initial Data

| Atribut Value | High import clusters | |
|------------------|------------------------|-----------|
| | Netto | CIF |
| | 8973929,9 | 3336257,3 |
| | Middle import clusters | |
| | Netto | CIF |
| | 1582007,8 | 463422,3 |
| | Low import clusters | |
| | Netto | CIF |
| | 3156,8 | 1779,5 |

By using the centroid can be clustered data that has been obtained into 3 clusters. Cluster process by taking the closest distance from any data that is processed. From the import data of rice of the leading country of origin, the clustering was obtained in the iteration of 1 to 3 clusters. High import level clusters (C1), i.e. Vietnam and Thailand, medium imported clusters (C2), i.e. China, India and Others and low imported cluster C3, i.e. Pakistan, USA, Taiwan, Singapore, and Myanmar. The process of finding the shortest distance, the data grouping in iteration 1 and Clustering data can be described in the following tables and figures:

Table 4. Calculation of cluster central gaps

| No | Country | Netto (Ton) | CIF (US\$) | C1 | C2 | C3 | Shortest Distance |
|----|-----------------------|----------------|---------------|---------|---------|---------|----------------------|
| 1 | Vietnam | 8973929,9 | 3336257,3 | 0 | 7930554 | 9570450 | 0 |
| 2 | Thailand | 5884429,9 | 2134647,2 | 3314948 | 4615607 | 6256077 | 3314948 |
| 3 | Tiongkok ¹ | 1582007,8 | 463422,3 | 7930554 | 0 | 1644957 | 0 |
| 4 | India | 1019299,9 | 305743,5 | 8512353 | 584382 | 1060632 | 584382 |
| 5 | Pakistan | 732306,3 | 234729 | 8805898 | 879939 | 765457 | 765457 |
| 6 | Amerika Serikat | 455790,5 | 154904,5 | 9092838 | 1167711 | 477833 | 477833 |
| 7 | Taiwan | 70968,5 | 20237,8 | 9500458 | 1574691 | 70279 | 70279 |
| 8 | Singapura | 3156,8 | 1779,5 | 9570450 | 1644957 | 0 | 0 |
| 9 | Myanmar | 212321,9 | 35435,1 | 9362756 | 1434996 | 211855 | 211855 |
| 10 | Lainnya | 851824 | 346917,4 | 8654753 | 739420 | 916164 | 739420 |

Table 5. Grouping of iterative data 1

| No | Country | C1 | C2 | C3 |
|----|-----------------------|----|----|----|
| 1 | Vietnam | 1 | | |
| 2 | Thailand | 1 | | |
| 3 | Tiongkok ¹ | | 1 | |
| 4 | India | | 1 | |
| 5 | Pakistan | | | 1 |
| 6 | Amerika Serikat | | | 1 |
| 7 | Taiwan | | | 1 |
| 8 | Singapura | | | 1 |
| 9 | Myanmar | | | 1 |
| 10 | Lainnya | | 1 | |

In table 5, the K-Means process will continue to iterate until the data grouping equals the previous iteration data grouping. In other words, the process will continue iterating until the data in the last iteration is the same as the previous iteration. The iterative grouping graph 1 can be seen in the following figure:

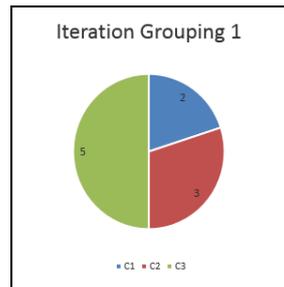
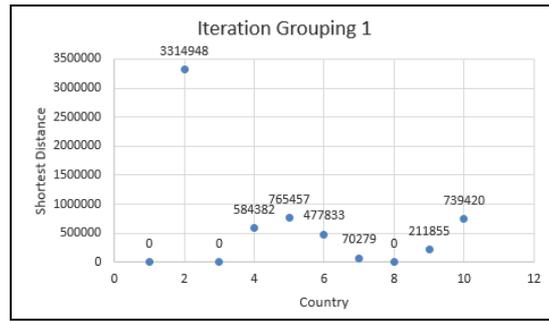


Fig. 1. Clustering of Iteration Data 1

In iteration 1, it is obtained clusters of rice import data of the main country of origin, which can be seen in Figure 2. In iteration 2, the process of midpoint or centroid value for iteration two can be seen in Table 6 below:

Table 6. Centroid Data iteration 2

| Atribut | High import clusters | |
|------------------------|----------------------|------------|
| | Netto | CIF |
| Value | 7429179,9 | 2735452,25 |
| Middle import clusters | | |
| | Netto | CIF |
| | 1151043,9 | 372027,733 |
| Low import clusters | | |
| | Netto | CIF |
| | 294908,8 | 89417,18 |

After getting the value of the midpoint or centroid, the same process is done by finding the closest distance. The process of finding the shortest distance, the data grouping in iteration 2 and Clustering data can be described in the following tables and figures:

Table 7. Calculation of the center distance of the iteration cluster 2

| No | Country | Netto (Ton) | CIF (US\$) | C1 | C2 | C3 | Shortest Distance |
|----|-----------------------|-------------|------------|---------|----------|----------|-------------------|
| 1 | Vietnam | 8973930 | 3336257 | 1657474 | 8365656 | 9266465 | 1657474 |
| 2 | Thailand | 5884430 | 2134647 | 1657474 | 5050918 | 5951950 | 1657474 |
| 3 | Tionggok ¹ | 1582008 | 463422,3 | 6273081 | 440548,4 | 1340337 | 440548,4 |
| 4 | India | 1019300 | 305743,5 | 6854929 | 147479,1 | 756002,3 | 147479,1 |
| 5 | Pakistan | 732306,3 | 234729 | 7148548 | 440672,3 | 460903,6 | 440672,3 |
| 6 | Amerika Serikat | 455790,5 | 154904,5 | 7435549 | 728367,9 | 173699,5 | 173699,5 |
| 7 | Taiwan | 70968,5 | 20237,8 | 7843192 | 1135922 | 234382,3 | 234382,3 |
| 8 | Singapura | 3156,8 | 1779,5 | 7913203 | 1206121 | 304630,3 | 304630,3 |
| 9 | Myanmar | 212321,9 | 35435,1 | 7705396 | 997243 | 98664,39 | 98664,39 |
| 10 | Lainnya | 851824 | 346917,4 | 6997622 | 300271,7 | 613564,1 | 300271,7 |

Table 8. Grouping of iterative data 2

| No | Country | C1 | C2 | C3 |
|----|-----------------------|----|----|----|
| 1 | Vietnam | 1 | | |
| 2 | Thailand | 1 | | |
| 3 | Tiongkok ¹ | | 1 | |
| 4 | India | | 1 | |
| 5 | Pakistan | | 1 | |
| 6 | Amerika Serikat | | | 1 |
| 7 | Taiwan | | | 1 |
| 8 | Singapura | | | 1 |
| 9 | Myanmar | | | 1 |
| 10 | Lainnya | | 1 | |

From table 8, the result of the 2nd iteration grouping is obtained with unequal results in the iteration 1. The process will take place in the next iteration. The iterative grouping graph 1 can be seen in the following figure:

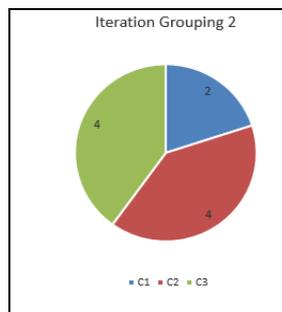
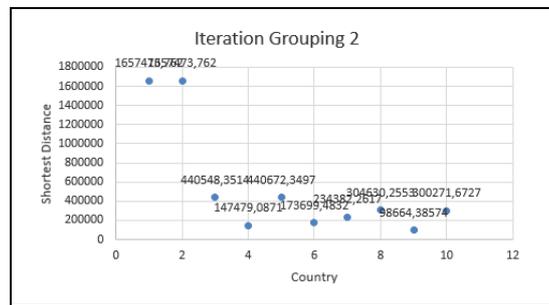


Fig. 2. Clustering of Iteration Data 2

Based on figure 2, from the import data of rice of the main country of origin, the clustering of iterations 2 for 3 clusters is obtained. High import level clusters (C1) ie Vietnam and Thailand, medium-level import clusters (C2) ie China, India Pakistan and Others and imported low-level clusters (C3) ie the United States, Taiwan, Singapore and Myanmar. The result of iteration 2 shows different results with the result of iteration 1. Then the process will proceed to iteration 3 with the same calculation that determines the new centroid based on iteration 2 and find the shortest distance from the centroid value on iteration 3

Table 9. Centroid Data iteration 3

| Atribut | High import clusters | |
|---------|------------------------|------------|
| | Netto | CIF |
| Nilai | 7429179,9 | 2735452,25 |
| | Middle import clusters | |
| | Netto | CIF |
| | 1046359,5 | 337703,05 |
| | Low import clusters | |
| | Netto | CIF |
| | 185559,425 | 53089,225 |

After getting the value of the midpoint or centroid, the same process is done by finding the closest distance. The process of finding the shortest distance, data grouping on iteration 3 and Clustering data can be described in the following tables and figures:

Table 10. Calculation of the center distance of the iteration cluster 3

| No | Country | Netto (Ton) | CIF (US\$) | C1 | C2 | C3 | Shortest Distance |
|----|-----------------------|-------------|------------|---------|----------|----------|-------------------|
| 1 | Vietnam | 8973930 | 3336257 | 1657474 | 8475712 | 9381612 | 1657474 |
| 2 | Thailand | 5884430 | 2134647 | 1657474 | 5161001 | 6067125 | 1657474 |
| 3 | Tionggok ¹ | 1582008 | 463422,3 | 6273081 | 550204 | 1455487 | 550204 |
| 4 | India | 1019300 | 305743,5 | 6854929 | 41876,42 | 871181,6 | 41876,42 |
| 5 | Pakistan | 732306,3 | 234729 | 7148548 | 330504,3 | 576129,5 | 330504,3 |
| 6 | Amerika Serikat | 455790,5 | 154904,5 | 7435549 | 618212,8 | 288775,3 | 288775,3 |
| 7 | Taiwan | 70968,5 | 20237,8 | 7843192 | 1025754 | 119206,9 | 119206,9 |
| 8 | Singapura | 3156,8 | 1779,5 | 7913203 | 1095955 | 189481,9 | 189481,9 |
| 9 | Myanmar | 212321,9 | 35435,1 | 7705396 | 887121,5 | 32060,85 | 32060,85 |
| 10 | Lainnya | 851824 | 346917,4 | 6997622 | 194753,6 | 728178,2 | 194753,6 |

Table 11. Grouping of iterative data 3

| No | Negara | C1 | C2 | C3 |
|----|-----------------------|----|----|----|
| 1 | Vietnam | 1 | | |
| 2 | Thailand | 1 | | |
| 3 | Tionggok ¹ | | 1 | |
| 4 | India | | 1 | |
| 5 | Pakistan | | 1 | |
| 6 | Amerika Serikat | | | 1 |
| 7 | Taiwan | | | 1 |
| 8 | Singapura | | | 1 |
| 9 | Myanmar | | | 1 |
| 10 | Lainnya | | 1 | |

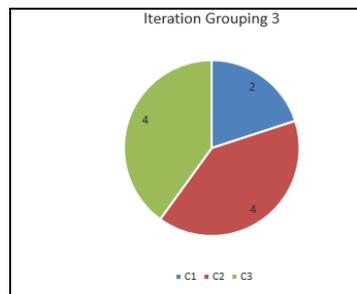
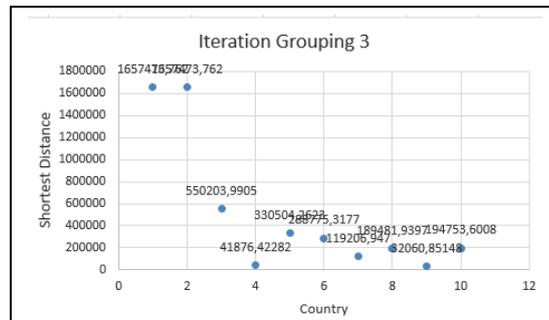


Fig. 3. Clustering of Iteration Data 3

In iteration 3, the grouping of data conducted on 3 clusters with iteration 2 obtained the same result. Of the 10 main rice import data of the leading country of origin can be known, 2 Cluster of high import level (C1) ie Vietnam and Thailand, 4 clusters of moderate import level (C2) ie China, India Pakistan and Others and 4 clusters of low import level (C3) United, Taiwan, Singapore and Myanmar.

IV. Conclusion

To assess the import of rice the leading country of origin can be Is done by applying a K-Means clustering method. The data is processed to derive the value of the imported rice production of the leading country of origin. The data were processed using Rapid miner to determine centroid values in 3 clusters, i.e., high introduced cluster level (C1), medium imported clusters (C2) and low imported level clusters (C3). Centroid data for imported high-level clusters 7429179,9 and 2735452,25, Centroid data for medium import level clusters 1046359.5 and 337703.05 and Centroid data for low import level clusters 185559.425 and 53089.225. Hence, it was obtained the assessment based on rice import index with 2 clusters of high import level (C1), i.e., Vietnam and Thailand, four medium imported clusters (C2), ie China, India Pakistan and Others and 4 clusters of low import level (C3), ie USA, Taiwan, Singapore, and Myanmar. The results of the research can be used to determine the amount of rice imported by the leading country of origin. In the processing of data for clustering can provide weighting criteria for data produced more accurate. Also, it also needs to increase the accuracy of clustering on the data. Added rules can be made to get more accurate results in the clustering process.

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