

Original Research Article

Evaluation of different clustering techniques in classifying the vegetable growing panchayats of Ernakulam district, Kerala

ABSTRACT

The goal of this study was to evaluate different clustering techniques in classifying the vegetable growing locations of Ernakulam (EKM) district of Kerala. Hierarchical clustering (HC) and K -means clustering were performed to group the panchayats based on soil fertility status and thereafter comparison of various clustering procedures was done using Davies – Bouldin (DB) index. Different dissimilarity measures- Euclidean, squared Euclidean, Chebychev distance and Mahalanobis D^2 were determined and single linkage, complete linkage and average linkage methods were adopted under these measures. The results revealed that Mahalanobis D^2 was the better clustering procedure with seven clusters (DB index: 0.120) followed by average linkage method under Euclidean distance (DB index: 0.306) with seven clusters. Manjapra and Keerampara panchayats remained as individual clusters. Keerampara had strongly acidic soils (pH -5.17) with high available Mg (158 mg kg⁻¹) while Manjapra soils had low Mg availability (19 mg kg⁻¹) and high S content (57 mg kg⁻¹). Kakkad, Kalady and Vengoor came under first cluster which possessed approximately same EC (0.15-0.19 dS m⁻¹), OC (2-2.4%) and Mg (71-73 mg kg⁻¹) content. Chengamanadu and Vengola came under third cluster while Ayyampuzha and Mudakkuzha belong to fourth cluster.

Keywords: Hierarchical clustering, Davies – Bouldin index, Mahalanobis D^2 , Euclidean distance, Average linkage.

1. INTRODUCTION

Kerala, the God's own country is blessed with biological diversity and soil heterogeneity. Ernakulam is a district situated in the central region of Kerala and rice is the major crop cultivated in the wet lands of the district. In addition to rice, vegetables, pineapple and nutmeg are also cultivated in EKM. Plant growth is highly dependent on the fertility status of soil prevailing in the region [1]. Plants need primary nutrients (Nitrogen (N), Phosphorus (P), Potassium (K)), secondary nutrients (Calcium (Ca), Magnesium (Mg), Sulphur (S)) and micro nutrients (Boron (B), Iron (Fe), Manganese (Mn), Zinc (Zn), Copper (Cu)) in adequate quantities for growth and development and they are absorbed from the soil. Other soil parameters viz. pH and Electrical conductivity (EC) also affect plant growth and nutrient availability [2].

Systematic testing of soil nutrients including the micro nutrients and prescribing recommendation is essential for soil health management. The soil test data available for each panchayat can be used for classifying them into various groups based on similarity in soil parameters. Cluster analysis, one of the multivariate methods is suitable in classifying objects based on similarity measures [3]. Clustering of panchayats based on their soil fertility status can be accomplished by means of cluster analysis [4]. Panchayats coming under the same cluster have similar soil characteristics and that under different clusters have dissimilar soil properties. Different clustering procedures are available which vary according to the distance measures selected for clustering. Same recommendations can be given to those panchayats which come under the same cluster.

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43 Cluster analysis is commonly applied in the field of plant breeding and genetics where
44 varieties/genotypes can be grouped based on their quantitative characters [5]. According to [6],
45 results of cluster analysis changes when different methods of clustering are used. Cluster analysis
46 can be used to partition sites based on soil characteristics [7]. The land use effect on soil chemical
47 and microbial properties can be determined with the use of cluster analysis [8]. Application of cluster
48 analysis on soil geochemical data helped to determine the spatial distribution of elements [9].
49 Evaluation of different clustering techniques is possible with a measure; Davies- Bouldin index which
50 identifies the cluster that best fit the data [10].

51 This study is an attempt to classify the panchayats of Ernakulam district based on the soil
52 fertility status. Among the various clustering procedures, clusters that best fit the data are also
53 identified.

54 2. MATERIALS AND METHODS

55 The present study is based on the data on thirteen soil fertility parameters of different vegetable
56 growing locations (panchayats) of Ernakulam district of Kerala. Soil samples collected from different
57 panchayats of Ernakulam district analysed by Department of Soil science & Agricultural Chemistry,
58 College of Agriculture, Vellayani, Kerala and the data maintained was utilized for the present study.
59 Sample collection was done by the farmers themselves from their own vegetable growing plots at a
60 depth of 15 cm using spade. From each sampling plot, about 10-15 random samples were collected,
61 mixed and reduced to 0.5 Kg by the method of quartering. It was observed that panchayats show
62 variation in soil properties with respect to cropping patterns and cultivation practices [11].

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63 The data on thirteen soil fertility parameters of 17 panchayats of Ernakulam viz., electro chemical
64 parameters (pH and Electrical conductivity (EC)), Oxidisable Organic Carbon (OC), Phosphorus (P),
65 Potassium (K), secondary nutrients (Calcium (Ca), Magnesium (Mg), Sulphur (S)) and micro nutrients
66 (Boron (B), Iron (Fe), Manganese (Mn), Zinc (Zn), Copper (Cu)) were available. Each panchayat is
67 having different sample sizes and altogether sample size comes around 583.

68 2.1 Cluster analysis (CA)

69 Cluster analysis is a multivariate technique used to group individuals or objects based on their
70 several characteristics [12]. There should be homogeneity within groups or clusters and heterogeneity
71 between groups. First we measure the distance between objects based on their multiple characters
72 which is otherwise called as similarity or dissimilarity measures. Based on the similarity/dissimilarity
73 measures, clusters are formed later.

74 2.1.1 Distance measures

75 Euclidean distance, squared Euclidean distance, Chebychev distance and Mahalanobis D^2
76 are the distance measures used for the present work.

77 2.1.1.1 Euclidean distance

78 It is the geometrical distance between two objects in the multidimensional space. It is
79 calculated as,

$$80 \quad d(x, y) = \sqrt{\sum_{i=1}^p (x_i - y_i)^2}$$

81 where X and Y are the 'p' dimensional vector of observations and $X = (X_1, X_2, \dots, X_p)$ and $Y =$
82 (Y_1, Y_2, \dots, Y_p) . Euclidean distance is one of the most commonly used distance measures.

83 2.1.1.2 Squared Euclidean distance

84 It is the square of Euclidean distance and is used to put weights for those objects which are
85 farther apart.

$$86 \quad E_{ij} = \sum_{k=1}^p (x_{ik} - x_{jk})^2$$

87 **2.1.1.3 Mahalanobis D^2 statistics**

88 A measure for group distance based on multiple characters was given by Mahalanobis, 1936. With $x_1,$
89 x_2, x_3, \dots, x_p as multiple measurements available on each individual and $d_1, d_2, d_3, \dots, d_p$ as $\bar{x}_1^1 - \bar{x}_1^2,$
90 $\bar{x}_2^1 - \bar{x}_2^2, \dots, \bar{x}_p^1 - \bar{x}_p^2$ respectively, being the difference in the means of two populations,
91 Mahalanobis D^2 statistics is defined as follows:

$$92 \quad D^2 = b_1 d_1 + b_2 d_2 + \dots + b_p d_p = \sum_i \sum_j (x_{i1} - x_{i2})' W^{-1} (x_{i1} - x_{i2})$$

93 Where \bar{x}_i^1 is the mean value of 1st character in the first population and \bar{x}_i^2 is the mean of the 1st
94 character in the second population. Here, the b_i values are to be estimated such that the ratio of
95 variance between the populations to the variance within the population is maximized. In terms of
96 variances and covariances, the D^2 distance between object 1 and object 2 can be obtained as;

$$97 \quad D^2 = (X_1 - X_2)' W^{-1} (X_1 - X_2)$$

98 Where W^{-1} is the inverse of variance covariance matrix, \bar{x}_1 is the mean of first population, \bar{x}_2 is
99 the mean of second population. It is used for quantitative data.

100 **2.1.1.4 Chebychev's distance**

101 This is a dissimilarity measure is based on the assumption that two objects are different if
102 they differ in any one of the characteristics and is calculated as,

$$103 \quad C_{ij} = \text{Max } |x_{ix} - x_{jx}|$$

104 **2.1.2 Clustering techniques**

105 These are the procedures used for clubbing together of similar objects into different clusters. There
106 are different techniques for clustering i.e. Hierarchical technique and K-means clustering [13].

107 **2.1.2.1 Hierarchical clustering**

108 Hierarchical clustering assumes each of the 'n' objects as individual clusters initially. Similar
109 objects are combined together in successive fusions or dissimilar objects are divided in successive
110 divisions in this clustering technique. There are different methods for linking the objects in different
111 clusters. Single linkage (nearest neighbour) method, complete linkage (farthest neighbour) method
112 and Tocher's method are common.

113 In single linkage, two individuals having minimum distance forms the first cluster. In next step, a third
114 individual is joined with the initial cluster or another two nearer individuals are clustered together to
115 form the second cluster. This is determined if the distance from the third individual to the first cluster is
116 shorter than the distance between the two nearer individuals. Two objects having maximum distance
117 between them constitute two groups. Next object either join one of the previous clusters or form its
118 own cluster.

119 Tocher's method of clustering objects makes use of Mahalanobis D^2 statistic. D^2 values are
120 arranged in ascending order and the two individuals having smallest distance between them is

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121 selected as the first cluster. Tocher suggested a cut off value which is equal to maximum among the
122 minimum D^2 values. Addition of a third object is determined in such a way that average D^2 distance is
123 less than the cut off value. Clustering continues until all the objects are included in one of the clusters.
124 Metode pengelompokan objek Tocher menggunakan statistik Mahalanobis D^2 .

125 2.1.2.2 K-means clustering

126 K-means clustering is a technique where we define some pre-specified number of clusters for
127 grouping 'n' objects. Only condition is that clusters are formed with minimum variance within clusters
128 and maximum variance between clusters.

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129 2.1.3 Cluster validity index

130 Davies-Bouldin index is one of the cluster validity measures to evaluate the clusters based on
131 their compactness and separation from each other. Let X_1, X_2, \dots, X_c are the clusters and $\Delta x_i, \delta(x_i, x_j)$
132 represents the intra cluster distance of i^{th} cluster and inter cluster distance between i^{th} and j^{th} cluster
133 respectively. Then DB index is defined as,

$$134 \quad DB = \frac{1}{c} \sum_{i=1}^c \max_{i \neq j} \frac{(\Delta x_i + \Delta x_j)}{\delta(x_i, x_j)}$$

135

136 DB index values should be less for good clustering i.e. small index values corresponds to clusters
137 which are compact and well separated. Clustering algorithms that minimize the DB index values give
138 optimum number of clusters.

139 3. RESULTS AND DISCUSSION

140 In order to classify the various panchayats based on the soil fertility status in Ernakulam
141 district, cluster analysis was adopted [14]. Different clustering algorithms such as K-means clustering
142 and hierarchical agglomerative clustering were adopted in the present study. Various dissimilarity
143 measures such as Euclidean distance, squared Euclidean, Mahalanobis D^2 and Chebychev distance
144 were estimated for clustering the panchayats. Various clustering procedures like single linkage,
145 complete linkage and average linkage were followed under hierarchical clustering to generate clusters
146 consisting of panchayats having similar soil fertility properties [15]. Tocher's clustering procedure was
147 also employed to generate clusters using Mahalanobis D^2 distance measure. Comparison of various
148 measures and clustering algorithms were completed using SPSS package. The study conducted by
149 [16] was in accordance with the present study as different clustering techniques provided different
150 number of clusters.

151 3.1 Hierarchical clustering of panchayats

152 3.1.1 Squared Euclidean distance

153 Squared Euclidean distance was selected for clustering the panchayats and the dissimilarity
154 matrix was determined. Single linkage, complete linkage procedures were practiced along with
155 average linkage method for squared Euclidean distance and the clustering of panchayats in each
156 case is given in Table 1.

157 There were seven clusters under single linkage and average linkage clustering procedure and
158 six clusters for complete linkage. Single linkage was another hierarchical clustering procedure and it
159 did not provide proper clusters of panchayats in EKM district..

160 It is evident from the Table 1 that Ayyampuzha, Chengamanadu, Keerampara, Thirumaradi
 161 and Thuravur formed as single clusters when single linkage was adopted. Five panchayats out of 17
 162 were in individual clusters. Only two clusters were having more than one panchayat in it and all other
 163 panchayats remained as individual clusters.

164 The panchayats Ayyampuzha, Chengamanadu, and Thuravur remained as single clusters
 165 under complete linkage and average linkage methods and Kadungalloor, Kalady and Puthenvelikkra
 166 were also in same cluster under these clustering procedures. There were six clusters under complete
 167 linkage and six clusters under average linkage. Ayyampuzha, Chengamanadu and Thuravur stood as
 168 individual clusters and there were only six clusters in complete linkage (Table 1).

169 **Table 1. Clustering of panchayats in EKM based on single linkage, complete linkage and**
 170 **average linkage method (Squared Euclidean distance).**

Cluster	Panchayats coming under each cluster		
	Single linkage	Complete linkage	Average linkage
I	Ayyampuzha	Ayyampuzha	Ayyampuzha
II	Kadungalloor, Kalady, Puthenvelikkara	Kadungalloor, Kalady, Puthenvelikkara	Kadungalloor, Kalady, Puthenvelikkara
III	Thuravur	Thuravur	Thuravur
IV	Chengamanadu	Chengamanadu	Chengamanadu
V	Keerampara,	Keerampara, Thirumaradi, Mudakkuzha, Nedumbassery, Piravom	Keerampara, Mudakkuzha, Thirumaradi
VI	Mudakkuzha, Kakkad, Vengola, Manjapra, Vengoor, Nedumbassery, Piravom, Pampakuda, Pothanikkad	Kakkad, Vengola, Manjapra, Vengoor, Pampakuda, Pothanikkad	Kakkad, Vengola, Manjapra, Vengoor
VII	Thirumaradi		Nedumbassery, Piravom, Pampakuda, Pothanikkad

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172 3.1.2 Euclidean distance

173 Euclidean distance was also selected as the distance measure and single linkage, complete
 174 linkage and average linkage procedures were performed under this distance measure. Average
 175 linkage using Euclidean distance as the distance measure gave the same clustering pattern as that of
 176 squared Euclidean (Table 2). Unlike squared Euclidean, single linkage and complete linkage using
 177 Euclidean distance resulted in six clusters and average linkage method had seven clusters.
 178 Ayyampuzha, Chengamanadu, Keerampara and Thuravur were formed as clusters with single
 179 panchayat in it and hence intra cluster distance was zero. Except Keerampara which was a single
 180 cluster for single linkage method only, Ayyampuzha, Chengamanadu and Thuravur were also found
 181 to be single clusters when complete and average linkage methods were used. Similar findings were
 182 put forwarded by [17].

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183 **Table 2. Clustering of panchayats in EKM based on single linkage, complete linkage and**
 184 **average linkage method (Euclidean distance).**

Cluster	Panchayats coming under each cluster		
	Single linkage	Complete linkage	Average linkage
I	Ayyampuzha	Ayyampuzha	Ayyampuzha
II	Thuravur	Thuravur	Thuravur
III	Chengamanadu	Chengamanadu	Chengamanadu
IV	Kadungalloor, Kalady, Puthenvelikkara	Kadungalloor, Kalady, Puthenvelikkara	Kadungalloor, Kalady, Puthenvelikkara
V	Keerampara	Keerampara, Mudakkuzha, Thirumaradi, Nedumbassery, Piravom	Keerampara, Mudakkuzha, Thirumaradi
VI	Nedumbassery, Piravom, Kakkad, Mudakkuzha, Pampakuda, Pothanikkad, Vengola, Vengoor, Manjapra, Thirumaradi	Kakkad, Vengola, Manjapra, Vengoor, Pampakuda, Pothanikkad	Kakkad, Vengola, Manjapra, Vengoor
VII		Kadungalloor, Kalady, Puthenvelikkara	Nedumbassery, Piravom, Pampakuda, Pothanikkad,

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187 Kadungalloor, Kalady and Puthenvelikkara came under a cluster when all these methods
 188 were practiced. This revealed that these three panchayats had almost similar soil characteristics.
 189 Similarly, Kakkad, Vengola, Manjapra and Vengoor were the panchayats which formed into a cluster
 190 and it was inferred that these panchayats also had similar soil properties.

191 Clusters obtained through Euclidean distance were similar as that of squared Euclidean distance.
 192 Ayyampuzha, Chengamanadu and Thuravur retained as individual clusters with intra cluster distance
 193 zero.

194 3.1.3 Mahalanobis D^2 distance

195 Panchayats coming under each cluster as per Mahalanobis D^2 distance is given in Table 3. This
 196 distance measure provided seven clusters when Tocher's clustering method was adopted. Manjapra
 197 and Keerampara were the panchayats which retained as single cluster with intra cluster distance zero
 198 when Mahalanobis D^2 was selected as the distance measure. This result is in line with the results of
 199 the research work done by [5] on clustering of rice genotypes.

200

201

202

203 **Table 3. Clustering of panchayats in EKM based on Tocher's method**

Cluster	Panchayaths coming under each cluster	204
I	Kakkad, Kalady, Vengoor	205
II	Kadungalloor, Pothanikkad, Puthenvelikkara, Pampakuda, Thirumaradi	206
III	Chengamanadu, Vengola	207
IV	Ayyampuzha, Mudakkuzha	208
V	Nedumbassery, Piravom	209
VI	Manjapra	210
VII	Keerampara	

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211 **3.1.4 Chebychev distance**

212 Clustering of panchayats based on average linkage, single linkage and complete linkage
 213 methods using Chebychev distance as the dissimilarity measure is given below in Table 4. There
 214 were 12 clusters into which panchayats were grouped when single linkage method was adopted.
 215 Other than Ayyampuzha, Chengamanadu and Thuravur which were separate clusters in all the
 216 clustering procedures, Manjapra, Vengoor, Thirumaradi, Puthenvelikkara, Mudakkuzha and
 217 Keerampara stood as individual clusters. Kadungalloor, Kalady and Puthenvelikkara were formed as
 218 a single cluster inferring that these three panchayats had almost similar soil nutrient status. Complete
 219 linkage and average linkage gave seven clusters. Mudakkuzha, Keerampara and Thirumaradi came
 220 under the same cluster when both average and complete linkage methods were adopted.

221 **Table 4. Clustering of panchayats in EKM based on Chebychev distance under different**
 222 **methods**

Cluster	Panchayats coming under each cluster			223
	Single linkage	Complete linkage	Average linkage	
I	Ayyampuzha	Ayyampuzha	Ayyampuzha	224
II	Thuravur	Thuravur	Thuravur	
III	Kadungalloor, Kalady	Kadungalloor, Kalady, Puthenvelikkara	Kadungalloor, Kalady, Puthenvelikkara	225
IV	Chengamanadu	Chengamanadu, Nedumbassery, Piravom	Chengamanadu	226
V	Mudakkuzha	Mudakkuzha, Keerampara, Thirumaradi	Mudakkuzha, Keerampara, Thirumaradi	227 228 229
VI	Vengola, Kakkad,	Vengola, Kakkad, Manjapra, Vengoor	Vengola, Kakkad, Manjapra, Vengoor	230
VII	Pothanikkad, Pampakuda, Nedumbassery, Piravom	Pothanikkad, Pampakuda	Pothanikkad, Pampakuda, Nedumbassery, Piravom	231 232
VIII	Vengoor			
IX	Puthenvelikkara			233
X	Keerampara			
XI	Manjapra			234
XII	Thirumaradi			235

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236 Chebychev distance also provided somewhat similar clustering patterns as that of squared Euclidean
 237 and Euclidean distance. Only difference was in the number of clusters formed when single linkage
 238 method used under the three distance measures. Chebychev distance gave 12 clusters while
 239 Euclidean and squared Euclidean gave 6 and 7 clusters respectively.

240 3.2 K-means clustering

241 K-means clustering defines the number of clusters in advance. In the present study, number
 242 of clusters was taken as 5, 6 and 7 in K-means clustering. When $k = 5$, $K = 6$ and $K = 7$ panchayats of
 243 Ernakulam was classified into five, six and seven groups respectively and are presented in Table 5. K
 244-means clustering mendefinisikan jumlah cluster terlebih dahulu.

245 Only Ayyampuzha was as an individual cluster for $K = 5, 6$ and 7 . Thuravur became a single cluster
 246 when $k = 6$. Nedumbassery, Piravom and Chengamanadu came under cluster II for $K = 5, 6$ and
 247 7 . Kadungalloor, Kalady and Puthenvelikkara belonged to same cluster which suggested that they had
 248 similar soil properties. The same clustering procedure was adopted by [7] for partitioning locations
 249 based on soil properties.

250 Thuravur, which was in the same cluster with other panchayats, now formed as a new individual
 251 cluster when $K = 6$ was selected. Panchayats were again redistributed among different clusters when
 252 K-means clustering with $K = 7$ were performed. Now, Ayyampuzha, Keerampara and Thuravur were
 253 in different clusters with intra cluster distance zero i.e. those cluster had single panchayats as the
 254 entity.

255 **Table 5. Clustering of panchayats in EKM based on K-means clustering**

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Cluster	Panchayats coming under each cluster			256
	K =5	K =6	K =7	257
I	Ayyampuzha	Ayyampuzha	Ayyampuzha	258
II	Chengamanadu, Nedumbassery, Piravom	Chengamanadu, Nedumbassery, Piravom	Chengamanadu, Nedumbassery, Piravom	259 260
III	Kadungalloor, Kalady, Puthenvelikkara	Kadungalloor, Kalady, Puthenvelikkara	Kadungalloor, Kalady, Puthenvelikkara	261 262 263
IV	Kakkad, Manjapra, Pampakuda, Pothanikkad, Thuravur, Vengola, Vengoor	Kakkad, Manjapra, Pampakuda, Pothanikkad, Vengola, Vengoor	Kakkad, Manjapra, Pothanikkad, Vengola, Vengoor	264 265 266
V	Keerampara, Mudakkuzha, Thirumaradi	Keerampara, Mudakkuzha, Thirumaradi	Mudakkuzha, Pampakuda, Thirumaradi	267 268
VI		Thuravur	Thuravur	269
VII			Keerampara	270

271

272 It was concluded that Ayyampuzha, Thuravur and Chengamanadu were the panchayats with different
 273 soil characteristics and did not form any group with other panchayats based on the soil fertility
 274 parameters. In most of the clustering procedures adopted, Kadungalloor, Kalady and Puthenvelikkara
 275 came under a cluster based on their soil properties which indicated that these three panchayats had
 276 almost similar soil properties..

277 **3.3. Cluster validity index- panchayats of Ernakulam**

278 Being unsupervised procedure, cluster analysis need evaluation of the results of different clustering
 279 procedures. Cluster validity means identifying the clusters that best fit to the given data. Davies-
 280 Bouldin index was one of such measures used for cluster validation which was used in the present
 281 study (Table 6). Comparatively low values are preferred for good clustering procedure in this method..

282 **Table 6. Cluster validity index for different clustering procedures (EKM)**

Distance measure	Linkage method	DB index score	
Squared Euclidean	Average linkage	0.412	283
	Complete linkage	0.427	284
Euclidean distance	Average linkage	0.306	285
	Complete linkage	0.894	286
Chebychev distance	Average linkage	0.458	287
	Complete linkage	0.383	288
Mahalanobis D^2		0.120	289
K – means clustering	K=5	0.566	290
	K=6	0.467	291
	K=7	0.497	292

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293
 294 Single linkage method was not considered as most of the panchayat was retained as single
 295 clusters with intra cluster distance zero. A comparison between average linkage and complete linkage
 296 was carried out using the D-B index [18]. D- B index should be less for the optimum clustering pattern
 297 and here Mahalanobis D^2 was found to be the best clustering measure followed by average linkage
 298 method with Euclidean as the distance measure. Cluster means were determined for Mahalanobis D^2
 299 and the results are given in Table 7..

300 **Table 7. Cluster mean based on Mahalanobi's D^2 (EKM)**

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Cluster no.	pH	EC	OC	P	K	Ca	Mg	S	B	Fe	Mn	Zn	Cu
I	5.297	0.144	1.959	47.067	298.562	530.051	71.953	22.364	0.546	85.713	26.012	3.673	2.805
II	4.880	0.129	1.668	49.522	191.996	520.996	29.114	28.579	0.743	60.605	17.956	2.122	4.346
III	4.640	0.197	1.573	86.173	231.798	414.339	55.541	24.920	0.906	35.649	12.252	1.349	4.266
IV	5.207	0.151	1.911	56.747	346.930	522.571	108.995	13.525	0.717	76.649	29.685	2.436	3.313
V	5.083	0.103	1.394	55.613	165.148	369.000	80.500	20.948	0.863	38.035	13.610	1.797	1.505
VI	5.284	0.284	1.789	82.774	209.016	490.526	19.684	57.653	1.076	115.790	27.500	1.621	2.011

VII	5.171	0.066	1.928	31.203	139.112	274.655	41.172	21.247	1.015	158.241	21.764	5.709	Comment [S19]: the table should not be broken
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301

302 Keerampara and Manjapra remained as separate panchayats when Mahalanobis D^2 was
303 practiced. Even though both panchayats come under the agro ecological unit 3.1 (southern and
304 central foot hills), they were not found together under a cluster as they would be having dissimilar soil
305 characteristics [19]. Keerampara soils were strongly acidic with comparatively lower available Ca
306 content (274 mg kg⁻¹) and high Fe status (158 mg kg⁻¹). Some parts of Keerampara had proximity
307 with water bodies. Nedumbassery and Piravom belonged to cluster V with EC (0.10 dS m⁻¹), OC
308 (1.39 %), K (165.14 kg ha⁻¹) and S (20.94 mg kg⁻¹). Cluster IV comprised Ayyampuzha and
309 Mudakkuzha panchayats with an average pH (5.2), EC (0.15), OC (1.91 %) and K (346.93 kg ha⁻¹).
310 Kadungalloor, Pothanikkad, Puthenvelikkara, Pampakuda and Thirumaradi clubbed together to form a
311 cluster with 1.6 per cent OC, P (49.5 kg ha⁻¹) and K (191.99 kg ha⁻¹). Kakkad, Kalady and Vengoor
312 came under cluster I as they had approximately the same EC (0.15-0.19 dS m⁻¹), OC (2-2.4%) and
313 Mg (71-73 mg kg⁻¹). Chengamanadu which was reported as an individual cluster in all the clustering
314 methods adopted, came along with Vengola under Tocher's method. OC ranged from 1.55 to 1.58
315 per cent, K (216.88-246.71 kg ha⁻¹), Mg (54-56 mg kg⁻¹) and B (0.89-0.91 mg kg⁻¹) in these
316 panchayats.

317 For all the clustering methods used except Mahalanobis D^2 , Ayyampuzha retained as a single
318 cluster. Hierarchical clustering also provided information that Ayyampuzha did not form any group
319 with other panchayats. Ayyampuzha had moderately acidic soils (pH - 5.8) with high available K (570
320 kg ha⁻¹) while Chengamanadu had high available P (113 kg ha⁻¹) and low Ca content (300 mg kg⁻¹).
321 Ayyampuzha comes under the agro ecological unit 4.1 (Southern high hills) and lies near to
322 waterbodies. Thuravur was deficient in available B (0.28mg kg⁻¹). Kadungalloor, Kalady and
323 Puthenvelikkara came under cluster IV and the soils in this cluster recorded with high K (342.82 kg
324 ha⁻¹) and Ca (619.82 mg kg⁻¹). The cluster V comprised of Keerampara, Mudakkuzha and
325 Thirumarady and the soils were strongly acidic with comparatively lower P (33.84 kg ha⁻¹) status. The
326 soils of panchayats in cluster VI were moderately acidic with sufficient quantity of soil nutrients.
327 Cluster means calculated for different clusters based on Euclidean distance with average linkage
328 method is presented in Table 8.

329 **Table 8. Cluster mean based on Euclidean distance (EKM)**

Cluster no.	pH	EC	OC	P	K	Ca	Mg	S	B	Fe	Mn	Zn	Cu
I	5.8 4	0.1 5	1.7 3	63.72	570.2 1	670.7 0	109.7 7	8.12	0.5 0	19.99	16.4 0	2.3 0	1.1 7
II	5.0 9	0.1 5	1.7 4	41.61	112.9 8	616.4 4	33.68	16.0 5	0.2 8	37.77	29.1 2	6.5 1	4.3 4
III	4.2 4	0.2 5	1.5 6	113.5 9	246.7 1	300.7 1	56.79	27.2 9	0.8 9	43.97	7.89	1.2 9	7.3 0
IV	4.9 9	0.1 7	1.7 2	71.44	342.0 1	619.8 2	41.81	25.0 6	0.6 4	85.70	15.2 1	1.7 8	5.9 2
V	4.6 8	0.1 4	2.1 6	33.84	102.8 7	336.0 0	60.32	28.4 9	0.9 9	134.7 3	24.8 9	3.1 3	4.7 5
VI	5.4 0	0.1 6	1.7 1	56.91	242.0 5	497.7 9	54.18	33.6 6	0.6 8	71.39	23.3 5	2.8 1	2.2 8
VII	5.2 2	0.1 5	1.5 2	58.70	217.0 7	470.0 1	57.12	26.7 7	0.8 0	70.42	21.2 8	2.5 7	2.0 2

330 Among the average and complete linkage methods under Euclidean and squared Euclidean
331 distances, clustering based on average linkage method provided better clusters as the DB index was
332 low. DB index of complete linkage under squared Euclidean was 0.427 which was higher than that of

333 average linkage (0.412). Average linkage method had small DB index (0.306) under Euclidean
334 distance compared to squared Euclidean.

335

336 4. CONCLUSION

337 Soils exhibit a high degree of heterogeneity with respect to place and climate. Spatial
338 distribution of soil nutrients also vary with locations [20]. Fertilizer recommendations are given based
339 on the soil test results of the locations and if the locations having similar soil characteristics can be
340 grouped based on soil fertility status, recommendations could be given easily [21].

341 Clustering of panchayats in Ernakulam were carried out using hierarchical clustering and *K*-means
342 clustering. Different distance measures like euclidean distance, squared Euclidean, Chebychev
343 distance and Mahalanobis D^2 along with different linkage methods – single, complete and average
344 linkage were used for clustering the panchayats based on soil fertility status. . Davies- Bouldin index
345 was determined for each clustering procedure and Mahalanobis D^2 was found to be the best
346 clustering method followed by euclidean distance with average linkage method as the D-B index was
347 small for Mahalanobis D^2 . Pengelompokan panchayats di Ernakulam dilakukan dengan menggunakan
348 hierarkis clustering dan *K*-means clustering.

349 Clustering of panchayats in Ernakulam based on Mahalanobis D^2 resulted in seven clusters.
350 Keerampara and Manjapra remained as individual clusters. Keerampara soils were strongly acidic
351 with EC value 0.06 dS m⁻¹. Available Ca was observed to be comparatively lowest (274 mg kg⁻¹)
352 while Fe content was the highest (158 mg kg⁻¹). Available Mg was very low in Manjapra (19 mg kg⁻¹)
353 compared to other panchayats. On the contrary, S content was more in this panchayat (57 mg kg⁻¹).
354 Nedumbassery and Piravom belonged to the same cluster due to the similarities in their soil
355 characteristics. There were same EC (0.10 dS m⁻¹), OC ranged from 1.2 to 1.5 per cent, K ranged
356 from 161 to 168 kg ha⁻¹ and S ranged from 20 to 21 mg kg⁻¹ in these two panchayats. Kakkad,
357 Kalady and Vengoor belonged to the same cluster as EC (0.15-0.19 dS m⁻¹), OC (2-2.4%) and Mg
358 (71-73 mg kg⁻¹) had approximately same values. Pengelompokan panchayat di Ernakulam

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