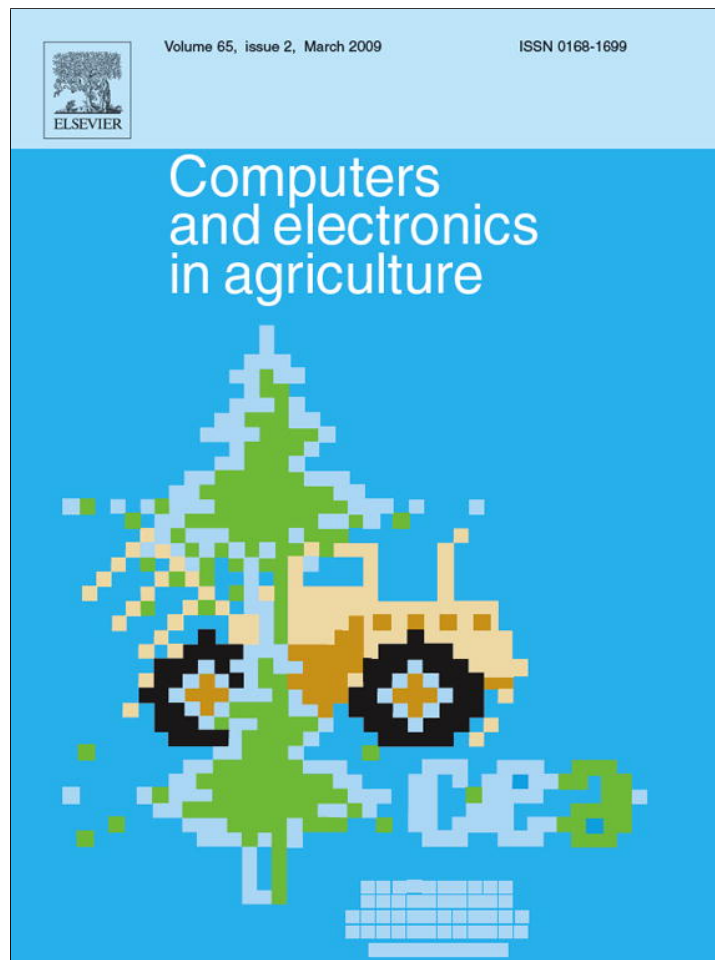


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Mahalanobis Taguchi System based criteria selection for shrimp aquaculture development

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ABSTRACT

Aquaculture development suitable location is not formulated from one criterion but from multiple criteria. A list of criteria was selected by reviewing the literature and consultation with aquaculture experts and then classified into five categories namely; water, soil, support, infrastructure and risk factor. Using these criteria a Mahalanobis Taguchi System (MTS) based tool was developed to facilitate the selection of prime set of criteria, which is a subset of the original criteria for suitable aquaculture site identification. Mahalanobis Distance (MD) was used for constructing a measurement scale for the aquaculture system. The principles of Taguchi method was used for screening the important criteria in the system. The information required to make an assessment was entered into the computer program. The computer program then processes the information and produces the prime set of criteria for each category. The software program is a valuable tool for the identification of prime set of criteria within each category. One can use these prime set of criteria within each category instead of all criteria for the identification of optimal location in shrimp aquaculture.

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1. Introduction

Aquaculture is emerging as a prime industry to tap the enormous turnover of bioenergy for the benefit of mankind. Close to 20% of all seafood consumed is now produced by aquaculture and global demand for aquaculture products was 22–24 million metric tonnes in the year 2000, a 14–24% increase since 1990 (Chamberlain and Rosenthal, 1995). India's present population is projected to hit 1.48 billion in the year 2030 leaving a wider gap in the supply demand stand (Brown et al., 1997). This gap has to be fulfilled with aquaculture production.

Uncontrolled expansion of aquaculture units and their intensive farm practices have brought severe stress on the surrounding environment. At present shrimp farming areas are rapidly expanding horizontally due to the increasing demand of shrimps in the international markets. This rapid expansion

of shrimp farming areas in the recent years is not following the ways of proper planning and management. A key factor in improving the environmental management of aquaculture is to develop methods of optimizing site selection for aquaculture.

The selection of optimal location in aquaculture system is not formulated just from one criterion but from multiple criteria (Ramesh and Rajkumar, 1996). Since all the criteria may not be necessary for the decision process, it is important to identify the prime set of criteria, which is a subset of the original criteria. In order to identify the prime criteria, experimental design becomes complex and difficult to manage. In this study, an alternative new approach is identified to the experimental design, the Mahalanobis Taguchi System (MTS) based decision tool. The purpose is MTS-based criteria selection for suitable shrimp farming area identification.

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2. Materials and methods

This section discusses the study area, decision criteria and its suitability ratings, normal and abnormal observations, and MTS framework. In presenting the framework, an outline of the method as used was provided.

2.1. Study area

The study area was Nellore district in the state of Andhra Pradesh, India. This district was selected because coastal aquaculture is a major economic activity in these areas. The Nellore district area is about 13,076 km². This district lies between the latitude of 13°37'–15°51' and the longitude of 79°15'–80°15' (Fig. 1).

2.2. Decision criteria and its suitability ratings

The key factors to be considered for selecting the optimal location in aquaculture are the availability of good quality water, soil quality, salinity, temperature, pollution, exposures to flood, infrastructural facilities and access to essential inputs and markets (Nath et al., 2000).

A list of criteria was selected by reviewing the literature and consultation with researchers who are all working aqua-

culture field. These criteria thus selected were then classified into five categories namely: water, soil, support, infrastructure and risk factor. The suitability ratings for the identified criteria for shrimp culture were drawn from different literatures with slight modification to suit the Indian environment based on the opinion of 25 researchers who are working in the aquaculture field. The group final decision was reached through majority (Guzzo, 1982). The criteria and its suitability ratings of shrimp are depicted in Table 1. The suitability ratings are given in three categories: (a) "Suitable", if the corresponding criterion favorable for designated use and no unusual construction, design, management or maintenance will be required in the identified site for the designated use; (b) "Moderately", if the corresponding criterion not favorable for designated use and special attention will be required in the identified site for the designated use; (c) "Severe", if the corresponding criterion unfavorable for designated use and the identified site may be useful for the designated purpose due to the major reclamation and modification in design, management or maintenance, but it may not be economically feasible to use a site with one or more unsuitable limitations.

2.3. Normal and abnormal observations

The data used in this study were collected from 52 randomly selected shrimp farms including normal and abnormal obser-

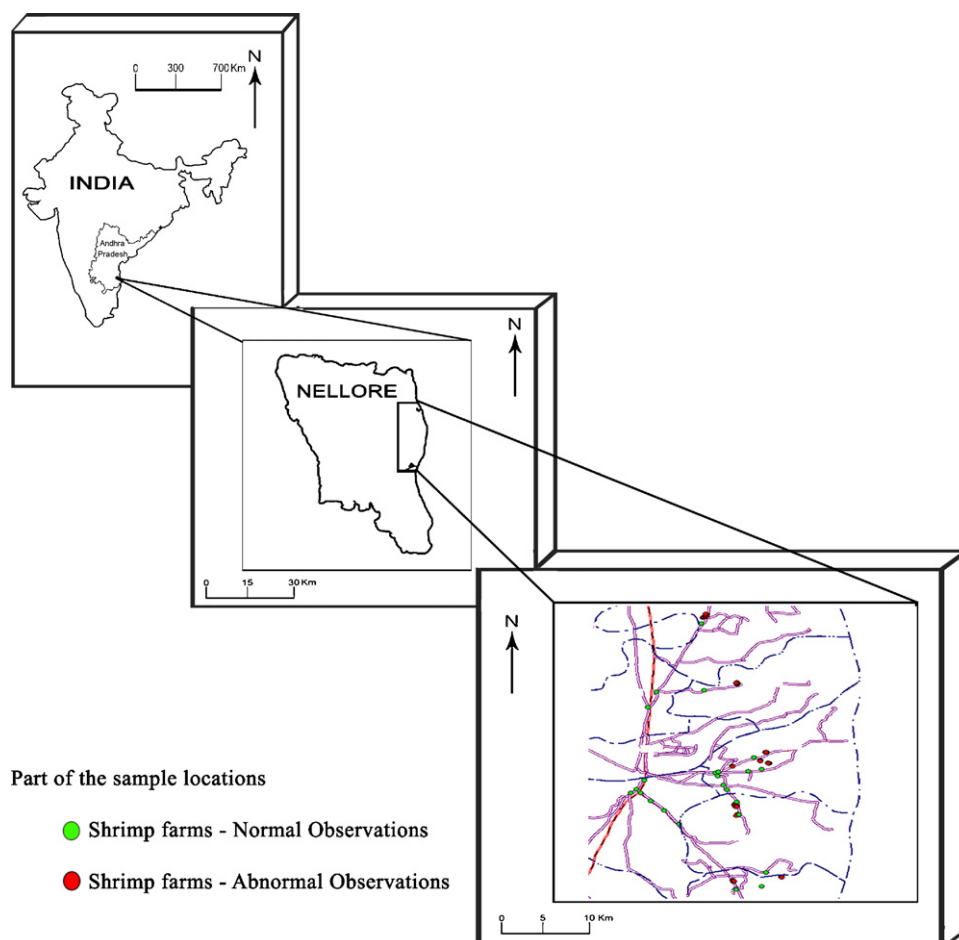


Fig. 1 – Location map.

Table 1 – Criteria (before and after application of MTS) and its suitability ratings of shrimp.

Sub-criteria (before MTS)	Suitability ratings			Notation for implementation	Sub-criteria (after MTS)
	Suitable	Moderate	Unsuitable		
<i>Water</i>					
Salinity (ppt)-Brackishwater	[15, 25]	(5, 15) or (25, 35)	≤5 or ≥35	W1	Salinity (ppt)-Brackishwater
pH (ppt)	[6.5, 8.5]	(5, 6.5) or (8.5, 10)	≤5.0 or 10.0	W2	pH (ppt)
Total alkalinity (mg/l)	[50, 200]	(20, 50) or (200, 500)	≤20 or ≥500	W3	Total alkalinity (mg/l)
Total hardness (mg/l)	[50, 200]	(20, 50) or (200, 500)	≤20 or ≥500	W4	Total hardness (mg/l)
Total hardness/total alkalinity	1	(0.5, 1)	≤0.5 or ≥	W5	Total hardness/total alkalinity
Dissolved oxygen (mg/l)	[5, 10]	(2, 5) or (10, 15)	≤2 or ≥15	W6	Dissolved oxygen (mg/l)
Free NH ₃ -N (mg/l)	≤0.1	(0.1, 1.0)	≥1.0	W7	Free NH ₃ -N (mg/l)
H ₂ S (mg/l)	[0, 0.1]			W8	H ₂ S (mg/l)
Temperature (°C)	[28, 33]			W9	Temperature (°C)
Transparency (cm)	[30, 60]	(15, 30) or (60, 120)	≤15 or ≥120	W10	Transparency (cm)
<i>Soil</i>					
pH (ppt)	[6.5, 9.0]	(4.5, 6.5)	≤4.5 or ≥9.0	S1	pH (ppt)
Salinity (ppt)	[8, 26]	(4, 8) or (26, 37)	≤4 or ≥37	S2	Salinity (ppt)
Clay content (%)	≥35	(18, 35)	≤18	S3	Clay content (%)
Organic carbon (%)	[1.5, 2.5]	(0.5, 1.5)	≤0.5 or ≥2.5	S4	Organic carbon (%)
Available nitrogen (mg/100 g)	[50, 75]	(25, 50)	≤25 or ≥75	S5	–
Available phosphorous (mg/100 g)	[4, 6]	(3, 4)	≤3 or ≥6	S6	Available phosphorous (mg/100 g)
<i>Support</i>					
Distance to NGOs (m)	≤5000	(5000, 15,000)	≥15,000	SU1	Distance to NGOs (m)
Distance to govt. offices (m)	≤3000	(3000, 10,000)	≥10,000	SU2	–
Distance to research station (m)	≤5000	(5000, 15,000)	≥15,000	SU3	–
Agglomeration (m)	≤1000	(1000, 3000)	≥3000	SU4	–
Distance to university/college (m)	≤5000	(5000, 15,000)	≥15,000	SU5	Distance to university/college (m)
<i>Infrastructure</i>					
Distance to natural fry (m)	≤3000	(3000, 12,000)	≥12,000	I1	Distance to natural fry (m)
Distance to processing plants (m)	≤3000	(3000, 10,000)	≥10,000	I2	Distance to processing plants (m)
Distance to rivers (m)	[500, 1000]	(1000, 4000)	≤500 or ≥4000	I3	–
Distance to roads (m)	≤2000	(2000, 5000)	≥5000	I4	–
Distance to local market (m)	≤1000	(1000, 3000)	≥3000	I5	Distance to local market (m)
<i>Risk factor</i>					
Flood and cyclone (per year)	0	1	≥1	R1	Flood and cyclone (per year)
Winter rain (m)	≤200	(200, 350)	≥350	R2	Winter rain (m)
Pollution (m)	≥4000	(2000, 4000)	≤2000	R3	Pollution (m)

vations (Fig. 1). The normal group, called Mahalanobis Space (MS), was constructed based on observations on 40 farms, which do not have the value of any criteria in the severe rating condition, that is, the value of criteria either in suitable or moderate ratings. The abnormal group was constructed based on 12 farms, which have the value of criteria in the severe rating condition.

2.4. Implementation

Matlab software (<http://www.mathworks.com>) was used to develop the MTS-based decision tool. The software has many in-built functions, such as transpose of matrix and inverse of matrix, which will be useful for the study. We need not

write the separate routine or module for the same and also the coding part is simple and understandable for the end user.

2.5. Mahalanobis Taguchi System

The selection of prime set of sub-criteria for each main criterion was computed iteratively using the following steps (Taguchi and Jugulum, 2000, 2002; Taguchi et al., 2001).

2.5.1. Construct a measurement scale with MS as the reference

The first step for the construction of measurement scale was the collection of normal observations and then normalized these observations using mean and standard deviation

obtained from the normal observations. Mahalanobis Distances (MDs) corresponding to all these observations were computed using inverse of correlation matrix method (Taguchi et al., 2001).

$$MD_j = D_j^2 = \frac{1}{k} Z_{ij}^T C^{-1} Z_{ij} \quad (1)$$

where Z_i is the normalized vector obtained by normalizing the values of X_i ($i = 1, 2, 3, \dots, k$)

$$Z_{ij} = \frac{X_{ij} - \bar{X}_i}{S_i}, \quad i = 1, 2, \dots, k; \quad j = 1, 2, \dots, n \quad (2)$$

$$\bar{X}_i = \frac{\sum_{j=1}^n X_{ij}}{n_i} \quad (3)$$

where X_{ij} is the value of i th sub-criteria in j th observation.

$$S_i = \sqrt{\frac{\sum_{j=1}^n (X_{ij} - \bar{X}_i)^2}{n - 1}} \quad (4)$$

where S_i is the standard deviation of i th sub-criteria, k is the number of sub-criteria, n is the number of observations, T is the transpose of the normalized vector and C^{-1} is the inverse of the correlation matrix.

It can be seen that MD in Eq. (1) was obtained by scaling (that is by dividing with k) the original MD. Since MDs were used to define the normal group we called this group as Mahalanobis Space. Mahalanobis Space is a database for the normal group consisting of mean vector, standard deviation vector, and inverse of the correlation vector (Taguchi and Jugulum, 2000). This Space provides a reference point for the measurement scale. According to MTS theory the average value of MDs is one for all the observations in MS (Taguchi et al., 2001).

2.5.2. Validating the measurement scale

The accuracy of the scale was justified by measuring the MDs of the known abnormal observations. The data from abnormal observations were normalized using mean and standard deviation obtained from the normal observations. The MDs were obtained for the abnormal observations using the correlation matrix of normal observations in Eq. (1). After calculating MDs, the average MD for the normal observations was compared with abnormal observations. The higher values of MDs for the

abnormal group validate the accuracy of the measurement scale.

According to the MTS theory, the MD of abnormal observations will be larger than the MD of normal observations if this is a good scale. Otherwise, one has to re-sample or find new sub-criteria to build the MS again.

2.5.3. Identification of the prime set of sub-criteria

In this step, prime set of sub-criteria was identified using orthogonal arrays (OAs) and signal-to-noise (S/N) ratios. The suitable OA was selected depending on the total degrees of freedom required to study the individual sub-criteria. The number of degrees of freedom is one less than the number of levels associated with the sub-criterion (Antony and Antony, 2001). The individual 'k' sub-criteria under study were assigned to the first 'k' columns of the identified OA (Taguchi, 1987). Level 1 in the identified OA column represents the presence of a sub-criterion and level 2 represents the absence of that sub-criterion. Inside an OA, every row represents the experimental combination of a run.

Using the sub-criteria combinations in the identified OA, MDs for the known abnormal observations were obtained using Eq. (1). From the MDs a larger-the-better S/N ratio was obtained for the q th run using the formula (Taguchi et al., 2001).

$$\frac{S}{N} \text{ ratio} = -10 \log_{10} \left[\left(\frac{1}{t} \right) \sum_{i=1}^t \frac{1}{MD_i^2} \right] \quad j = 1, 2, \dots, t \quad (6)$$

where t is the number of sub-criteria present for a given combination of the experimental run.

An average S/N ratio was calculated for each sub-criterion at level 1 and level 2. Subsequently, gain in S/N ratio values were calculated as:

$$\text{Gain} = \left(\text{average of } \frac{S}{N} \text{ ratio} \right)_{\text{level 1}} - \left(\text{average of } \frac{S}{N} \text{ ratio} \right)_{\text{level 2}} \quad (7)$$

If the gain is positive, keep the sub-criterion; if not then exclude it, for the next step.

2.5.4. Confirmation run

A confirmation run was conducted with the prime set of sub-criteria identified from the previous step. The reduced

Table 2 – MDs range of normal and abnormal group for each main criterion.

MDs (range and average)	Water	Soil	Support	Infrastructure	Risk factor
<i>Normal</i>					
Range	0.3655–1.4305	0.4962–1.7949	0.1002–2.8624	0.1471–2.2818	0.2709–3.02796
Average	0.9475	0.9583	0.9436	0.9410	0.9977
<i>Abnormal (all sub-criteria)</i>					
Range	54.9894–325.8896	11.7431–27.913	5.4683–196.5499	5.9123–12.9733	9.9476–25.3095
Average	165.28	18.8027	100.9772	9.2736	16.7423
<i>Abnormal (prime set of sub-criteria)</i>					
Average	85.9437	21.58344	244.7575	12.4950	7.6588

Table 3 – $L_{12}(2^{11})$ OA and average response for larger-the-better S/N ratios (water).

Run	1 (W1)	2 (W2)	3 (W3)	4 (W4)	5 (W5)	6 (W6)	7 (W7)	8 (W8)	9 (W9)	10 (W10)	11
1	1	1	1	1	1	1	1	1	1	1	1
2	1	1	1	1	1	2	2	2	2	2	2
3	1	1	2	2	2	1	1	1	2	2	2
4	1	2	1	2	2	1	2	2	1	1	2
5	1	2	2	1	2	2	1	2	1	2	1
6	1	2	2	2	1	2	2	1	2	1	1
7	2	1	2	2	1	1	2	2	1	2	1
8	2	1	2	1	2	2	2	1	1	1	2
9	2	1	1	2	2	2	1	2	2	1	1
10	2	2	2	1	1	1	1	2	2	1	2
11	2	2	1	2	1	2	1	1	1	2	2
12	2	2	1	1	2	1	2	1	2	2	1
Level 1	15.3424	15.5052	15.9630	15.3200	15.7005	15.6429	16.0740	15.8727	16.0590	14.2493	
Level 2	14.3802	14.2174	13.7596	14.4026	14.0221	14.0797	13.6485	13.8499	13.6636	15.4733	
Gain	0.9621	1.2878	2.2033	0.9175	1.6785	1.5632	2.4255	2.0228	2.3954	-1.2240	

measurement scale was constructed using the prime set of sub-criteria identified and then MDs corresponding to the abnormal observations were obtained using Eq. (1). In the next step, the average MD based on the abnormal observations obtained with prime set of sub-criteria identified was compared to that obtained from all the sub-criteria originally used.

If the average MD of abnormal group with prime set of sub-criteria identified was lower than that of all the sub-criteria, retain the excluded sub-criteria also in the prime set of sub-criteria identified; if not then consider only the sub-criteria which were identified as prime set from the previous step, for the prediction of future observations.

Table 4 – $L_8(2^7)$ OA and average response for soil, support, and infrastructure.

Run	1	2	3	4	5	6	7
1	1	1	1	1	1	1	1
2	1	1	1	2	2	2	2
3	1	2	2	1	1	2	2
4	1	2	2	2	2	1	1
5	2	1	2	1	2	1	2
6	2	1	2	2	1	2	1
7	2	2	1	1	2	2	1
8	2	2	1	2	1	1	2
	Sub-criteria						
	(S1)	(S2)	(S3)	(S4)	(S5)	(S6)	
Soil							
Level 1	8.3538	8.0795	10.5344	8.2307	7.4301	8.1429	
Level 2	7.3371	7.6114	5.1565	7.4602	8.2608	7.5480	
Gain	1.0166	0.4682	5.3778	0.7705	-0.8307	0.5949	
	Sub-criteria						
	(SU1)	(SU2)	(SU3)	(SU4)	(SU5)		
Support							
Level 1	13.2238	11.6154	11.3544	11.8034	13.6409		
Level 2	10.7311	12.3395	12.6005	12.1515	10.3141		
Gain	2.4926	-0.7241	-1.2461	-0.3480	3.3268		
	Sub-criteria						
	(I1)	(I2)	(I3)	(I4)	(I5)		
Infrastructure							
Level 1	9.7749	8.6041	7.5980	7.3057	9.5403		
Level 2	7.2316	8.4024	9.4084	9.7008	7.4661		
Gain	2.5433	0.2017	-1.8104	-2.3951	2.0742		

Table 5 – $L_4(2^3)$ OA and average response for larger-the-better S/N ratios (risk factor).

Run	1 (R1)	2 (R2)	3 (R3)
1	1	1	1
2	1	1	1
3	1	2	2
4	1	2	2
Level 1	10.9501	10.1701	8.2293
Level 2	6.9392	7.7192	9.6599
Gain	4.0109	2.4509	–1.4306

3. Results and discussion

With the help of 40 normal observations, the MDs corresponding to all these observations were computed for each main criterion. Mahalanobis Space was defined for each main criterion with the help of the MDs obtained for 40 observations. The data from the 12 abnormal observations were normalized using mean and standard deviation obtained from the 40 normal observations. The MDs corresponding to all these observations were estimated for each main criterion. Table 2 gives the MDs range of normal and abnormal group with their corresponding average value for each main criterion. It was seen that the average MD for the normal observations for each main criterion was close to the theory of MTS. Since the MDs of abnormal for each main criterion have higher distances, the measurement scale was considered as good.

There are many orthogonal arrays in Taguchi's method; however, in this study, $L_{12}(2^{11})$ (Table 3), $L_8(2^7)$ (Table 4), and $L_4(2^3)$ (Table 5) OAs were selected for the identification of the prime set of sub-criteria from their corresponding main criteria with minimum number of sub-criteria combinations. In Tables 3–5, "1" (or level 1) represents including the sub-criterion and "2" (or level 2) excluding the sub-criterion. An average S/N ratio was obtained for each sub-criterion at level 1 and level 2. Subsequently, gain in S/N ratio was calculated by the difference between the average S/N ratio value at level 1 and level 2 (Tables 3–5).

From Table 3, it is clear that in water the nine sub-criteria such as: W1, W2, W3, W4, W5, W6, W7, W8 and W9 have positive gains. That means these sub-criteria have higher average responses when they are part of the system (level 1). Hence these sub-criteria were considered to be useful for confirmation run. A similar interpretation was true for the rest of the main criteria in Tables 4 and 5.

A confirmation run was conducted for each main criterion with their corresponding prime set of sub-criteria identified. MDs corresponding to the normal and abnormal observations were obtained using Eq. (1). Comparisons of average MD for normal and abnormal group for each main criterion were given in Table 2. It can be observed from Table 2 that all the main criteria, except water and risk factor, have higher value of average MD based on the abnormal group (obtained with only their corresponding prime sub-criteria identified as useful) compared to that obtained from all the sub-criteria originally

used. This shows that, we only need to collect the samples for soil, support, and infrastructure from their corresponding prime set of sub-criteria, to identify the optimal location in aquaculture. The results also revealed that, in water and risk factor, the abnormal have lower average MD with their corresponding prime set of sub-criteria combinations. Therefore, the sub-criterion transparency and pollution cannot be reduced from water and risk factor respectively for future observations (Table 2). Table 1 shows the results of MTS algorithm, that is, the required set of sub-criteria in each main criterion for aquaculture development.

4. Conclusion

A simple-to-use MTS-based decision tool that assists for the selection of important criteria which will be useful for the identification of optimal location for aquaculture development has been developed. One can use the MTS-based tool in aquaculture system without the knowledge of statistics. If the discrimination is performed alone by an individual criterion, it may produce a misleading result. The proposed MTS decision tool combines all criteria into an MD index with considering the correlation among all criteria. The proposed decision tool can be easily adapted to other closely related industries such as agriculture and environmental engineering. This tool is also useful for demonstration and training purposes.

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