Outliers Detection on Fisheries Commodity Transaction from Local Market in Tual City based on The x-means Clustering

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Abstract. Correctly estimating factors that affect the income and profitability of coastal communities specifically catching fishermen is one way to increase the level of sustainable coastal community welfare. The purpose of this study is to detect outliers from fisheries commodity transactions in the fish market in Tual City, to have full view on which cases that may have certain traits that can improve the profitability of fishermen in the region. The research method used in this study is a survey method in which used as many as 1420 datasets obtained from interviews conducted starting from June 2017 to June 2018. The data mining approach is used in this study, in order to obtain the appropriate outliers then x-means clustering is used to cluster profitable transaction data, then data collected further being process with outlier detection. We found 25 high profitability transactions of 1420 datasets. The outlier analysis was carried out and found 1 outliers. We have figured that the species of fish and the location of the seller are the factors that influence the level of profit from the fishermen. This result is an input for policy makers and catch fishermen in conducting capture fisheries.

1. Introduction

Increasing the dynamics of development in an area is highly dependent on an effort to optimize the use of existing resources in the region for the prosperity of the people and for the benefit of all components in the region. Optimization of resource utilization in an area will depend on the knowledge of those who utilize the potential of natural resources In Maluku province there are three fisheries management areas (WPP), respectively: WPP-714 Banda Sea, WPP-715 Arafura Sea and WPP-716 Seram Sea, Maluku Sea and Tomini Bay. The Kei Islands is one of the regencies and cities in Maluku, which is between two WPPs that needs good capabilities and good knowledge. The potential of capture fisheries in the Kei Islands and Maluku Province in general according to [1] is still a source of income for fishermen in this region but refers to the existing conditions and trends related to behavior, fleets and techniques including illegal fishing so the tendency of the development of marine resources in particular fish will decrease in the next 8-10 years. The Kei Islands are included in the coral triangle waters region, so it is important to maintain the rich biodiversity of marine resources in the waters.

Based on studies by [2] showed that a total of over IDR 240 million was generated on average by each of the coastal villages in the Kei Islands, including the coastal villages in Tual City. This value is
still below an optimal number based on the daily needs of coastal communities and the population of coastal communities added with given the characteristics of coastal communities in this region which has a very high number of family dependents with an average low level of education [3]; [4]; [5]; [6]. This condition is compounded by the financial and capital capability to procure better fishing gear. This causes income from capture fishermen in this region to be very inconsistent, one of the causes which also one cliché problem of coastal communities in the Kei Islands is a result of the use of technology that is still limited and limited knowledge [5]. Another factor that is an obstacle to improving the welfare of coastal communities is the potential for conflicts that arise both between villages, in villages between families in the village due to the use of marine resources [6]; [7]. For this reason, research is needed that not only improves the efficiency of fish catching activities, but also avoids potential conflicts and increases in income of coastal communities. A data mining approach can be used to help resolve this problem.

Data mining is process where an intelligent method are applied to extract knowledge and often associated with term knowledge discovery [8]. The data mining approach has been widely used by various fields including in the fields of fisheries and marine affairs [9]; [10]; [11]; [12]. Utilization of this technique is very helpful in breaking down information and knowledge from the data obtained. To increase fish catch efficiency by fishermen and increase fishermen profits, it can be obtained by identifying fish species, catch time and fishing location. Thus, the purpose of this study is to detect outliers from consumer transaction datasets on fisheries commodities in the local fish market in Tual City, Kei Islands, which provide the highest profit value to obtain the empirical characteristics of these transactions to obtain a fishing profile that benefits fishermen in the area.

2. Methodology
The stages of this research can be seen in Fig. 1, where the data mining approach used in this study consists of two parts, namely cluster analysis using the x means clustering approach which then clusters with the highest profits will be used to be analyzed for outliers to identify transactions that are outside trend and has a unique profile, namely transactions that provide high profits with fish species, catch time and unique fishing locations. The data used in this study were 1420 datasets obtained from a survey for one year in May 2017-June 2018 on fish commodity transactions in the local fish market in Tual City.

![Figure 1. Research framework](image)

2.1. X-means clustering
X-means clustering is a variation of K-means clustering, x-means clustering is a variation of K-means clustering. The x-means clustering algorithm is summarized as follows [13]:

1. Set initial number of clusters \( k_0 \), then applied k-means algorithm into all data \( (k_0 = k) \) and repeat all procedure (setting \( i = 1, 2, \ldots, k_0 \)).
2. Apply k-means cluster with setting \( k = 2 \) for cluster \( C_1, C_2, \ldots, C_{k_0} \).
3. Assume \( p \)-dimensional distribute normal for data \( x \) in \( C_1 \). Calculate BIC as, \( BIC = -2 \log L(\hat{\theta}_i; x \in C_1) + q \log n_i \) and \( BIC' \) as, \( BIC' = -2 \log L'(\hat{\theta}_i'; x \in C_1) + q' \log n_i \). Where, \( \hat{\theta}_i \) is the maximum likelihood estimation of \( p \)-dimensional, \( q \) is number of dimensions of the parameters and becomes \( 2p \), \( n_i \) is the number of elements contained \( C_1 \) and \( L(.) \) is likelihood function, \( q' \) is given by \( 2 \times 2p = 4p \).
4. When \( BIC > BIC' \) continue the division and if the vice versa, \( BIC < BIC' \), divide further clusters and stop.
5. Number the cluster output identification from the center of each cluster, and the number of elements in each cluster.

### 2.2. Outliers detections

Anomaly detection identifies single records in datasets which significantly deviate from the normal data, further it is process of finding patterns in a given dataset which deviate from the characteristics of the majority. We can use the maximum likelihood method to estimate the parameters \( \mu \) and \( \sigma \) by maximize the log-likelihood function, as follows [8]:

\[
\ln \mathcal{L}(\mu, \sigma^2) = \sum_{i=1}^{n} \ln f(x_i | (\mu, \sigma^2)) = -\frac{n}{2} \ln (2\pi) - \frac{n}{2} \ln \sigma^2 - \frac{1}{2\sigma^2} \sum_{i=1}^{n} (x_i - \mu)^2
\]

Where \( n \) is the total number of samples, for derivatives \( \mu \) and \( \sigma^2 \), solving the first order condition leads to maximum likelihood estimates:

\[
\hat{\mu} = \bar{x} = \frac{1}{n} \sum_{i=1}^{n} x_i
\]

\[
\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^{n} (x_i - \bar{x})^2
\]

### 3. Result and Discussion

#### 3.1. Clustering analysis

The results of data analysis using the x-means clustering approach found that there were 4 clusters, each of which was divided based on the main criteria, namely the level of income received from fishermen. Measurement of cluster distance using euclidean distance with the average distance of each cluster is 0.950 with Davies-Bouldin Index of 0.752. In Fig 2, it appears that cluster 3 is a cluster that is used to the next stage for detecting outliers. Obtained 25 transaction data from 1420 datasets which have a price of only 5.97% higher than other datasets, but have an earning rate of 604.59% greater than other datasets and a higher sales amount of 517.73%.


**Figure 2. x-means clusters summary**

**Table 1. Cluster 3 datasets**

| No | Year | Period (month) | Seller Origin | Fish’ Species | Number of selling | Price | Earnings | Fishing Ground |
|----|------|----------------|----------------|---------------|-------------------|-------|----------|----------------|
| 1  | 2017 | June           | Fair           | Lethrinidae   | 143               | 50.000 | 7,150,000 | Langgiar      |
| 2  | 2017 | June           | Warbal         | Stolephorus   | 300               | 10.000 | 3,000,000 | Debut         |
| 3  | 2017 | July           | Mangon         | Sardinella maderensis | 430 | 10.000 | 4,300,000 | Kaimana       |
| 4  | 2017 | July           | Komp Banda Ely | Auxis rochei | 270               | 30.000 | 8,100,000 | Banda Ely     |
| 5  | 2017 | July           | Ngursit        | Auxis rochei  | 145               | 50.000 | 7,250,000 | Ngursit       |
| 6  | 2017 | July           | Sathean        | Sardina pilchardus | 290 | 20.000 | 5,800,000 | Ohoilili      |
| 7  | 2017 | July           | Sathean        | Sardina pilchardus | 195 | 20.000 | 3,900,000 | Ohoilili      |
| 8  | 2017 | July           | Banda Ely      | Auxis rochei  | 150               | 25.000 | 3,750,000 | Banda Ely     |
| 9  | 2017 | July           | Banda Ely      | Auxis rochei  | 203               | 25.000 | 5,075,000 | Banda Ely     |
| 10 | 2017 | July           | Banda Ely      | Moolgarda seheli | 165 | 25.000 | 4,125,000 | Kaimana       |
| 11 | 2017 | July           | Wara           | Auxis rochei  | 211               | 25.000 | 5,275,000 | Ngursit       |
| 12 | 2017 | July           | Wara           | Auxis rochei  | 213               | 25.000 | 5,325,000 | Kei Besar     |
| 13 | 2017 | July           | Un             | Sardina pilchardus | 500 | 20.000 | 10,000,000 | Pulau Ut      |
| 14 | 2017 | July           | Fair           | Auxis rochei  | 39                | 30.000 | 1,170,000 | Ngursit       |
| 15 | 2017 | July           | Banda Ely      | Auxis rochei  | 150               | 25.000 | 3,750,000 | Banda Ely     |
| 16 | 2017 | August         | Wara           | Katsuwonus pelamis | 70 | 50.000 | 3,500,000 | Perairan      |
| 17 | 2017 | August         | Fair           | Auxis rochei  | 315               | 20.000 | 6,300,000 | Ngursit       |
| 18 | 2017 | August         | Mangon         | Rastrelliger faughni | 150 | 25.000 | 3,750,000 | Dobo          |
| 19 | 2017 | August         | Un             | Stolephorus   | 60                | 10.000 | 600,000   | Lupus         |
| 20 | 2017 | August         | Banda Ely      | Auxis rochei  | 15                | 50.000 | 750,000   | Ngursit       |
| 21 | 2017 | August         | Kiom           | Parexocoecus brachypterus | 150 | 20.000 | 3,000,000 | Perairan      |
| 22 | 2017 | August         | Fidabot        | Ocyurus chrysurus | 20 | 50.000 | 1,000,000 | Tam           |
| 23 | 2017 | August         | Fidabot        | Stolephorus   | 70                | 10.000 | 700,000   | Lupus         |
| 24 | 2017 | August         | Fair           | Auxis rochei  | 200               | 20.000 | 4,000,000 | Ngursit       |
| 25 | 2018 | April          | Kompleks Banda | Selar crumenophthalmus | 60 | 20.000 | 1,200,000 | Sathean       |
3.2. Outliers detection
Based on the results of the x-means cluster analysis that has been done in the previous stage, then it has been used to detect outliers. Thus, it is expected to produce outliers that have unique characteristics and are out of trend to further serve as a reference for fishermen and fishery commodity sellers in Tual City in running related businesses. In Fig 3a, a comparison of fish species and fishing ground is seen that has unique characteristics. It is seen that on Langgiar fishing ground which is located in Kei Besar Selatan Barat sub-district, Southeast Maluku district with fish species that being capture was Lethrinidae. Then when seen in Fig. 3b, information on catch periods has been obtained in June which is included in The Indian summer monsoon. Thus, both fish species and period of catching information have been found, which it has the potential to provide more benefits to fishermen and sellers in terms of earning and profit.

![Graph showing fish species and fishing ground](image)

**Figure 3.** Outliers detection: (a) combination between period of catching and fish species; (b) combination between fishing ground and fish species

4. Conclusion
Based on the results of research that has been done, x-means clustering gained 25 datasets out of 1420 datasets of fishery commodity transactions in the local fish market in Tual City which has high profit characteristics. Further results were obtained that, there were 1 data outliers from 25 datasets included
in cluster 3. From the information of the outliers, it is obtained that the benefits obtained by the data are 57% higher than the average datasets in cluster 3, with a selling price of 53% higher than the average price in cluster 3. The results also found two factors that differentiate the fish species, Lethrinidae and the fishing period in June. Implication of research is that fishermen in this area should pay attention to fish species that are caught and sold because it is very influential on the income earned, this is related to taste of customer and distance of fish capture. In addition, fishers and seller also have to put more attention to period of fishing (monsoon), so it could be more efficient considering the natural and marine conditions in the Kei Islands, this is related to the demand and availability of fish in the market.

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