

Analysis and Estimation of Crop Yield Using DBSCAN Algorithm

Latha G¹, Anusha S¹, Vikas B O²

¹B.E Students, Department of ISE, New Horizon College of Engineering, Bangalore, Karnataka, India

²Assistant Professor, Department of ISE, New Horizon College of Engineering, Bangalore, Karnataka, India

ABSTRACT

Agriculture field requires farmers and agriculture businesses to take critical decisions every day and also overcome the complexity and factors that they influence. Approach for accomplishing practical and effective solutions for this problem. Agriculture has been an significant target for big data. The various factors such as production, rainfall, temperature, soil type, fertilizer used, influence the farmer to access information and to make critical farming decision. This paper focuses on the analysis of the agriculture data and finding optimal parameters to maximize the crop production using data mining techniques like DBSCAN and Multiple Linear Regression. The result is the yield estimation and this helps agribusinesses to make future business decisions.

Keywords: Clustering, DBSCAN, Linear Regression, Agricultural data

I. INTRODUCTION

Today, India holds second position in agriculture. Agriculture plays a very country. There are many factors that influence agriculture some of them are soil, important role and is a backbone of our cultivation, irrigation, fertilizers, temperature, rainfall, harvesting, pesticides also important for farmers and companies engaged in industries[5]. Industries that are weeds, production and other factors. Historical data of crop yield information is based on food, paper, and organic products depend on agriculture for crops. An accurate and optimal estimation of crop production, risk these companies in planning supply chain decision like production scheduling.

Data mining plays a vital role in data analysis and is one of the growing field. Data mining is a process of analysing large amount of data and finding pattern and predicting the future outcome which is required.

Data mining contains many different methods for mining the data. One such technique is clustering. Clustering is the process of grouping common data into clusters or data that belong to same class based on the constraint. Clustering has various methods like

- ✓ K-means
- ✓ Hierarchical
- ✓ Density based clustering
- ✓ K-mediod

Cluster analysis divides data into well-formed groups. Well-formed clusters should capture the “natural” structure of the data. The proposed method focuses on DBSCAN clustering methods.

II. PROPOSAL

A. Problem Formation

The existing system is implemented using k-means clustering algorithm. It is a simple clustering algorithm. The disadvange in the existing system is

that the clusters formed are only in spherical shape[4]. The number of clusters to be formed on agricultural data must be passed initially in the code. It is very difficult to predict the K value. The cluster formed contains noise and is not accurate.

B. Problem Solution

The project mainly focuses on estimating the crop yield which is more accurate and optimal, noise free that helps farmers and agricultural business to make critical decisions. The overall system design of the project is shown below.

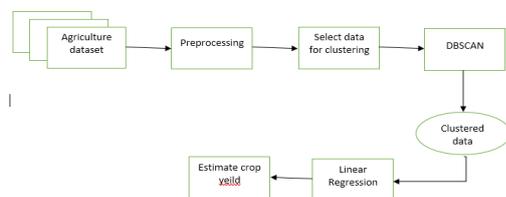


Figure 1. shows system design

The data set used in our project contain the attributes like fertilizer, area available for cultivation, temperature, cereals production, oilseeds production,, pulses production, pulses yield, oilseeds yield, cereals yield.

C. Data Preprocessing

The agricultural data contains missing values and noise points, to remove this we need to pre-process the data. Pre-process is also a process to obtaining the required data for further use[3], the result of pre-process is shown below.

```

120.97 97.01 122.84 59.93 119.58 124.21 122.2 84.9
15.32 97.01 123.11 97.01 115.32 79.45 116.96 100.04
1 123.53 105.85 120.97 124.21 120.97 105.85 123.11
  
```

Figure 1.2. is preprocessed data

D. DBSCAN Algorithm

The result of pre-processing is given as input to DBSCAN algorithm. Density-based spatial clustering of applications with noise (DBSCAN) is a data clustering algorithm. DBSCAN requires two parameters: ϵ (eps) that is radius and the minimum

number of points required to form a dense region (minPts) within which the cluster will be accepted. It starts with an arbitrary starting point that has not been visited. This point's ϵ -neighbourhood is retrieved, and if it contains sufficiently many points, a cluster is started. Otherwise, the point is labelled as noise. Note that this point might later be found in a sufficiently sized ϵ -environment of a different point and hence be made part of a cluster[7].

If a point is found to be a dense part of a cluster, its ϵ -neighborhood is also part of that cluster[6]. Hence, all points that are found within the ϵ -neighborhood are added, as is their own ϵ -neighborhood when they are also dense. This process continues until the density-connected cluster is completely found.

The algorithm can be expressed in pseudo code as follows:

```

AdvancedDBSCAN(DB, eps, minPts) {
  C = 0 /* Cluster counter */
  for each point P in database DB {
    if label(P) ≠ undefined then continue /* Previously
    processed in inner loop */
    Neighbors N = RangeQuery(DB, P, eps) /* Find
    neighbors */
    if |N| < minPts then { /* Density check */
      label(P) = Noise /* Label as Noise */
      continue
    }
    C = C + 1 /* next cluster
    label */
    label(P) = C /* Label initial point
    */
    Seed set S = N \ {P} /* Neighbors to
    expand */
    for each point Q in S { /* Process every
    seed point */
      if label(Q) = Noise then label(Q) = C /* Change
      Noise to border point */
      if label(Q) ≠ undefined then continue /* Previously
      processed */
      label(Q) = C /* Label neighbor */
    }
  }
}
  
```

```

Neighbors N = RangeQuery(DB Q, eps) /* Find
neighbors */
if |N| ≥ minPtsthen { /* Density check */
    S = S ∪ N /* Add new
neighbors to seed set */
} }
}

```

Where RangeQuery can be implemented using a database index for better performance, or using a slow linear scan:

```

RangeQuery(DB, Q, eps) {
    Neighbors = empty list
    for each point P in database DB { /* Scan all
points in the database */
        if dist(Q, P) ≤ eps then { /* Compute
distance and check epsilon */
            Neighbors = Neighbors ∪ {P} /* Add to
result */
        }
    }
    return Neighbors
}

```

The epsilon value in our project given is 5 and minimum number of points is 4. we can change the value of epsilon and minpts the result will be seen in cluster formation.

III. IMPLEMENTATION AND RESULT

```
AdvancedDBSCAN(dbscan, 5, 4) {
```

```

Cluster=0
Foe each point 97.01 in dbscan{
If 97.01 is not visited continue{
Neighbour N=RangeQuery(dbscan,97.01,5)

```

The output of the above example is a cluster3 for point 97.01.

```

[Instance86 115.32 96.91][Instance89 116.96 97.01][Instance20 116.96 97.01][Instance73 116.96 96.91][Instance255 116.7 97.01][Instance251 116.96 97.01]
[Instance90 116.96 97.01][Instance15 116.96 97.01][Instance88 116.96 97.01][Instance78 116.96 96.91][Instance59 116.96 97.01][Instance795 116.96 97.01]
[Instance94 116.7 96.91][Instance98 116.7 96.91][Instance50 116.96 96.91][Instance191 116.7 96.91][Instance891 116.7 96.91][Instance91 116.7 96.91]
[Instance96 116.7 96.91][Instance302 116.7 96.91][Instance70 115.32 96.91][Instance705 116.32 97.01][Instance94 116.7 97.01][Instance25 115.32 97.01]
[Instance33 116.7 97.01][Instance94 115.32 97.01][Instance23 116.7 97.01][Instance68 116.7 96.91][Instance77 116.7 97.01][Instance58 115.32 97.01]
[Instance69 115.32 97.01][Instance47 115.32 97.01][Instance47 115.32 97.01][Instance897 115.32 96.91][Instance6 115.32 97.01][Instance104 115.32 97.01]
[Instance88 116.7 96.91][Instance69 116.7 96.91][Instance53 116.7 96.91][Instance219 115.32 96.91][Instance12 115.32 96.91][Instance167 115.32 96.91]
[Instance81 115.32 96.91][Instance72 116.96 97.01][Instance83 116.96 97.01][Instance43 116.96 97.01][Instance205 116.96 97.01][Instance102 116.96 97.01]
[Instance45 116.96 97.01][Instance45 116.96 97.01][Instance42 116.96 97.01][Instance40 116.96 97.01][Instance85 122.2 96.91][Instance106 122.2 96.91]
[Instance92 122.2 97.01][Instance142 122.2 97.01][Instance32 122.2 97.01][Instance148 122.2 96.91][Instance79 122.2 96.91][Instance85 122.2 96.91]
[Instance33 122.2 97.01][Instance187 122.2 97.01][Instance388 122.2 97.01][Instance658 122.2 96.91][Instance58 122.2 96.91][Instance11 122.2 96.91]
[Instance76 122.2 96.91][Instance334 122.2 96.91][Instance52 122.2 96.91][Instance50 122.2 96.91][Instance83 122.2 96.91][Instance143 120.13 96.91]
[Instance84 120.13 96.91][Instance282 120.97 97.01][Instance34 120.97 97.01][Instance712 120.97 97.01][Instance470 120.97 97.01][Instance246 120.97 97.01]
[Instance89 120.97 96.91][Instance155 120.97 97.01][Instance207 120.13 97.01][Instance78 120.13 97.01][Instance268 120.13 97.01][Instance208 120.13 97.01]
[Instance7 120.97 97.01][Instance152 120.97 96.91][Instance126 120.97 97.01][Instance965 120.97 96.91][Instance847 122.2 96.91][Instance182 120.97 96.91]
[Instance757 120.97 96.91][Instance405 120.97 96.91][Instance56 120.97 96.91][Instance967 120.13 96.91][Instance304 120.13 97.01][Instance890 120.13 97.01]
[Instance93 120.13 97.01][Instance138 120.13 97.01][Instance343 120.13 97.01][Instance228 123.93 97.01][Instance689 120.13 97.01][Instance796 120.13 96.91]
[Instance754 120.13 96.91][Instance34 123.93 97.01][Instance121 124.32 97.01][Instance728 123.93 97.01][Instance68 123.11 97.01][Instance768 122.84 97.01]
[Instance643 122.84 97.01][Instance63 123.93 96.91][Instance481 124.32 96.91][Instance26 123.93 97.01][Instance301 124.32 96.91][Instance101 124.32 96.91]
[Instance770 124.32 96.91][Instance756 123.93 97.01][Instance51 123.93 97.01][Instance209 123.93 96.91][Instance76 123.93 96.91][Instance31 123.93 96.91]
[Instance508 123.93 96.91][Instance19 123.93 97.01][Instance519 123.93 96.91][Instance70 123.93 97.01][Instance10 123.93 97.01][Instance51 123.93 96.91][Instance62 123.93 96.91]
[Instance298 123.93 96.91][Instance41 123.93 96.91][Instance170 123.93 97.01][Instance10 123.93 97.01][Instance324 123.11 97.01][Instance17 123.11 96.91]
[Instance107 123.93 96.91][Instance507 122.84 97.01][Instance612 123.11 96.91][Instance324 123.11 97.01][Instance351 123.11 96.91][Instance66 123.11 96.91]
[Instance87 123.11 96.91][Instance764 123.11 97.01][Instance376 123.11 97.01][Instance435 123.11 96.91][Instance351 123.11 96.91][Instance66 123.11 96.91]

```

Figure 2. shows cluster 3

The result is the cluster formed and number of outliers detected. The clusters are formed based on the pre-processed data for example let's consider the dataset fertilizer and production from the pre-processed dataset the output obtained is 7 clusters with two noise points. The value in cluster may be repeated but all the points are visited. Clusters contain similar data of production and fertilizer. The two noise points are the outlier has they may have high value or does not fall in any of these clusters.

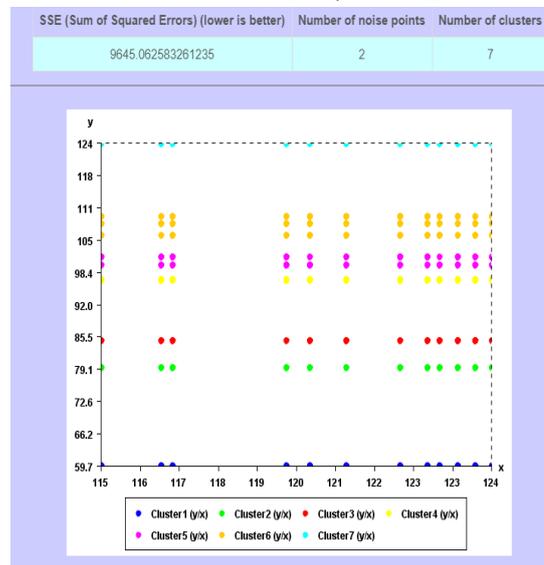


Figure 3. shows total clusters formed for the given dataset

The above figure shows the clusters formed. The output of algorithm is 7 clusters and two outliers. The clusters are dense. They can be repeated for other

attributes such as cereal production, fertilizer, yield etc.

Multiple Linear Regression

These cluster are the input to multiple linear regression. Multiple linear regression contains many independent variables and one dependent variable. Every value of the independent variable x is associated with a value of the dependent variable y . The linear equation in multiple linear regression shows how dependant values influence the output[8].

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 + R_2 X_4 + E$$

Beta0 is the y intercept, beta1 means single unit change in X_1 then Y changes beta1 times, beta2, beta3 and R2 are other independent values for area, temperature and production. The E is the error rate as there contains a little noise i.e. no data is perfect. The E value used it -1 in our project.

Beta values are the mean of the attribute of the particular attribute. It is very helpful because when data is more we can estimate our output using this mean values.

beta0	beta1	beta2	beta3	R2
1.2304207093046716	-12.114028246384681	6.0767967250006685	-0.3103267682529166	-0.013568166600796524

Figure 4. shows the beta values

Based upon the independent and the depended values we will be calculating the beta values. Depending upon the beta values, we will be estimating the crop yield. We will giving the current values for the fertilizer used, area, temperature, production and based upon the calculated beta values we will be estimating the yield.

The below figure shows the final yield estimated for the given values of fertilizer ,area, temperature and productction.The input values given are 100kg of

fertilizer,100 acres of land ,27 celsius temperature and production is 97.the output is 1496 quintels of yield.



Figure 5. shows the estimated yield for the given input

III. CONCLUSION

The DBSCAN algorithm forms clusters that are denser and noise free, the clusters can be of any shape. Through Linear regression we can get a yield that is more optimal and accurate. This helps the farmers and agribusiness to make future decisions.

As the future work we intend to expand of our dataset that contains more attributes like soil and other information of agriculture.

IV. REFERENCES

- [1]. "A survey on pre-processing and post processing techniques in data mining." Tomar, Divya, and Sonali All;arwal International Journal of Database Theory & Application 7.4 (2014).
- [2]. "Enhancing data analysis with noise removal." Xiong, Hui, et al, IEEE Transactions on Knowledge and Data Engineering 18.3 (2006): 304- 319.
- [3]. "Survey of Pre-processing Techniques for Mining Big Data" JayaramHariharakrishnan*, Mohanavalli.S*, Srividya*, Sundhara Kumar K.B** Department of Information Technology, SSN

College of Engineering Kalavakkam, Tamil Nadu, India.IEEE[2017].

- [4]. "Analysis of agriculture data using data mining techniques: application of big data" Jharna Majumdar* , Sneha Naraseeyappa and Shilpa Ankalaki [2017]
- [5]. "Demand Based Crop Recommender system for Farmers" S.Kanaga Suba Raja, Rishi .R, Sundarsan.E,Srijit .V Department of Information Technology , Easwari Engineering College, Chennai, India.[2017]
- [6]. "Convex-Hull & DBSCAN Clustering to Predict Future Weather" Ratul Dey , Sanjay Chakraborty Computer Science & Engineering ,Institute of Engineering & Management[2015]
- [7]. <https://sites.google.com/site/dataclusteringalgorithms/density-based-clustering-algorithm>
- [8]. <https://newonlinecourses.science.psu.edu/stat501/node/283/>