

Crop Yield Maximization Using an IoT-Based Smart Decision

¹Amita Shukla, ²Dr. Krishna Kant Agrawal

Submitted: 03/02/2024

Revised: 11/03/2024

Accepted: 17/03/2024

Abstract: This paper conducted a comprehensive analysis of the integration between sensor technologies and machine learning algorithms in terms of crop yield prediction for precision agriculture. Appreciating the role of precision yield estimate in mitigating global food challenges, this paper discusses various sensor technologies including NPK sensors among others; their strengths and weaknesses are highlighted. An in-depth analysis of machine learning algorithms such as Decision Trees, Naïve Bayes, Support Vector Machines, K-Nearest Neighbors and Ensemble Learning reveal their comparative performances with regards to adopting them into agricultural practices. In addition, the use of Multiple Linear Regression for planning rainfall enables an interdisciplinary approach to precision agriculture based on both soil characteristics and climatic conditions. The discussion covers the emerging trends, patterns and gaps in previous research evidence on this topic along with possible implications for future studies or concrete implementation. Through identifying the challenges and limitations, including periodic sensor calibration as well as algorithm interpretability furthered by the review our complex reality of precision farming.

Keywords: *Fack news, RandomForest, J48, SMO, NaiveBayes, OE-MDL, Ibk, LSTM .*

1. Introduction

Predicting crop yield is a key function that drives modern agriculture, which today uses it as the major tool for optimal resource use towards productiveness and risk reduction in an environment where conditions are uncertain. Farmers have numerous challenges that influence their endeavors in achieving most extreme edit generation. Climate designs, soil and natural variety as well as changes of bugs and infections are impressive dangers to trim generation. In any case, the ancient ways of cultivating that depend on encounter and instinct are now not sufficient to ensure nourishment security as well as financial victory. To neutralize these dangers, innovation gets to be an irreplaceable accomplice for agriculturists sharp on moving forward their decision-making methods. Innovation advancements in sensor innovations, information analytics strategies and machine learning algorithms offers an opportunity to convert agribusiness by giving a more profound understanding of soil quality as well as climate perspectives at the same time trim wellbeing. One approach of specific note is the utilize of ML methods that analyze huge information sets, joining soil values and climate markers and historical agricultural hones (Gao *et al.* 2023). Such developments make it conceivable to make predictive models that give ranchers with decision-oriented data necessary for way better

choices all through the method of edit development. This audit points to assess the existing inquire about on edit abdicate expectation by centering on the combination of sensor innovations and machine learning calculations. Through such an appraisal of current approaches' qualities and shortcomings, we trust to build up designs, crevices as well as regions for improvement within the given field. This evaluation moreover subtle elements the proposed "Selection of Edit Machine Learning Model" (SCS) and its outfit learning strategy that tosses a few light on how this demonstrate handles complex cultivating challenges. Finally, we look for to give our input into the existing discourse around utilizing innovation for advancing ecologically mindful and beneficial agrarian forms with present day security concerns in intellect.

II: Literature Review

Sensor Technologies

The utilize of a wide extend of sensor advances for measuring the soil parameters has been basic in exactness horticulture and trim surrender expectation by making a difference to secure exact, opportune data. These sensors incorporate distinctive advances that point to identify basic characteristics of the soil environment. One common shape is the coordinates soil sensor, which can analyze basic constituents such as nitrogen (N), phosphorous (P), potassium K, pH levels T , stickiness H and electrical conductivity EC. These sensors which are more often than not combined in one gadget give a all-encompassing viewpoint on the soil state, an vital prerequisite for edit development. As discussed by Goliński *et al.* (2023), More

¹Research Scholar, School of Computer Science and Engineering, Galgotias University, Greater Noida, India. Email: amitapandeybit@gmail.com

²Professor SCSE, School of Computer Science and Engineering, Galgotias University, Greater Noida, India. Email: kkagrwal@outlook.com

particularly, specific sensors just like the DHT11 sense temperature and mugginess whereas the MQ 135 sensor recognizes gasses such as smelling salts, nitrogen oxides as well as carbon dioxide. It is additionally through these sensors that they come together to create a composite appraisal of the soil wellbeing and natural components. It is additionally a major advantage that the sensor innovations permit one to obtain information in real-time, permitting agriculturists to create opportune choices. Consolidating a few sensors gives a comprehensive picture of the soil, which makes a difference in exact supplement administration and water system strategies. But these innovations are not idealize either. Different components such as sensor float, calibration issues and impedances with the environment may impact levels of exactness. As defined by Ahmadzadeh *et al.* (2023), Another jump to its broad selection has been the establishment and support costs, especially for cash-strapped ranchers. Striking a fine adjust between the qualities and shortcomings of these sensor advances is fundamental to completely infer their benefits in making strides rural hones.

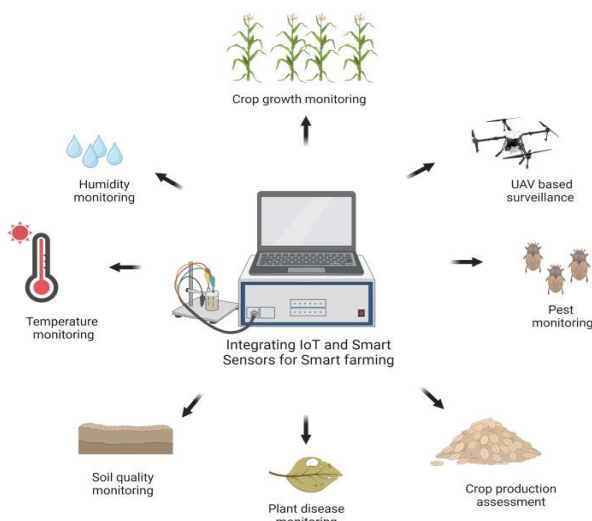


Fig 1: Sensor Technologies

Machine Learning Algorithms

The review paper employments an arms stockpile of ML calculations for trim abdicate determining, illustrating a advanced reaction to the multifaceted nature inalienable in farming. In specific, the Versatile K-Nearest Centroid Neighbor (aKNCN) classifier is utilized in soil quality examination that recognizes tests agreeing to their soil traits. As stated by Ahmed (2023), The moment level comprises the Extraordinary Learning Machine calculation and gathering learning strategies such as Choice Tree, Naïve Bayes, Back Vector Machine (SVM), K-Nearest Neighbor (KNN), and Arbitrary Timberland (RF) to foresee trim abdicate. Comparison of execution appears

that Irregular Woodland and Naïve Bayes calculations allow distant better;a much better;a higher;a stronger;an improved">a higher precision, accuracy esteem as well as review values in foreseeing reasonable crops. Arbitrary Woodland is characterized by tall stability and exactness, which permits it to be especially compelling within the foundation of choice boundaries. Naïve Bayes, as a probabilistic classifier illustrates viability in preparing different soil parameters. SVM and KNN algorithms perform sensibly well, hence contributing to the gathering execution. ML calculation appropriateness within the rural space is affected by components such as dataset estimate, computational effectiveness, and interpretability. The gathering nature of Irregular Timberlands makes them strong in managing with shifted and loud rural datasets. As explained by Sharma *et al.* (2022), Gullible Bayes being a likelihood show is in adjust with the dubious nature of agribusiness. On the one hand, computational complexity ought to be adjusted against benefits. The gathering strategy combining different calculations utilizes each algorithm's qualities, giving a all encompassing arrangement for precise trim surrender forecast over diverse rural settings

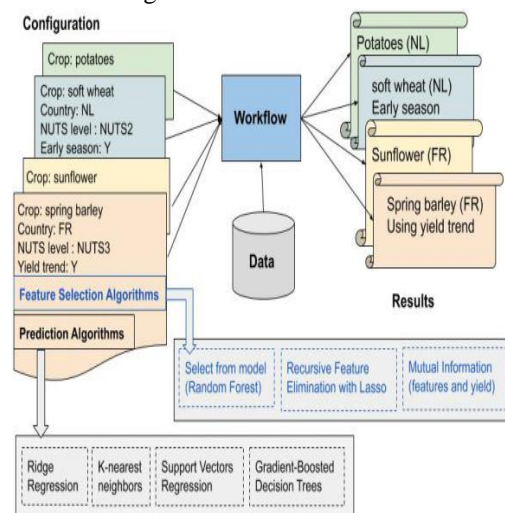


Fig 2: Machine Learning Algorithms

Ensemble Learning

The looked into paper employments one of the foremost capable standards – outfit learning, which gets to be an imperative calculate that makes strides expectation precision in trim surrender estimations. Gathering learning is based on combining expectations from many machine learning models to urge distant better;a much better;a higher;a stronger;an improved">an improved and more exact result than each of the person models. An imperative advantage is the minimization of overfitting, which it a infamous issue with machine learning. The combination of distinctive models permits outfit learning to relieve the chance of memorizing noise in preparing information, which leads to more generalized indicators. As discussed

by Ali *et al.* (2023), Also, gathering strategies counting the Arbitrary Woodlands utilized within the paper are prevalent at capturing complex designs and connections inside rural datasets. Gathering learning utilizes the assorted models' aggregate insights, where each show brings its claim interesting experiences to the table. This differing qualities makes a difference the gathering to generalize well across numerous scenarios, making strides show vigor and steadiness. Hence, gathering learning works as a durable approach that makes strides expectation precision and strength within the precarious zone of edit abdicate determining.

Rainfall Prediction

As for trim surrender expectation, exact precipitation estimating is significant to rural decision-making. Based on Different Straight Relapse (MLR), the surveyed paper is utilized to figure precipitation, which plays an critical part in trim administration. MLR uses past rain records to foresee future precipitation designs. Exact precipitation estimate engages ranchers to plan and water system plans, select fitting crops for development , as well as arrange the gather cycle in like manner. In zones where the need of rain is an issue, exact expectations on downpours permit agriculturists to satisfactorily utilize water preservation methods. As mentioned by Amertet *et al.* (2023), Furthermore, anticipating overwhelming downpours goes a long way to diminishing dangers related with immersion and soil misfortune. Coordination precipitation forecast into the edit abdicate demonstrate empowers the surveyed paper to offer a more cohesive approach, accuracy horticulture. The interaction between rainfall forecasting, and crop yield projections enables data-driven decisions thus promote sustainable agricultural systems.

Comparison with Existing Models:

As shown below, the SCS model provides a number of major strengths over existing precision agriculture approaches in terms of improved crop yield prediction and assistance to farmers' decision-making. A notable strength is the implementation of Ensemble Learning (EL) that essentially yields a collection of Decision Tree, Naïve Bayes, Support Vector Machine, K-Nearest Neighbor and Random Forest algorithms with predictive power. This combination increases the accuracy of prediction, which is a way to compensate for the flaws found in previous works by applying separate models. As defined by Assous *et al.* (2023), The other strength is the large dataset used for SCS model training, which includes 220 observations across ten features vital to crop maximization. This exceeds the dataset size of some other models, thus leading to better generalization and stability. How the model can be applied in various agricultural situations is further enhanced with diverse soil type such as loamy and clay. Nevertheless, it is important to note some shortcomings. However, the cost of

computing performed by Ensemble Learning can be a concern for environments with limited resources that affect real-time use. As defined by Balram & Kumar (2022), On the one hand, despite the model's accuracy advantage, it is necessary to note that in many cases of more general application or for broader selection of crops when there are significant difference between regions with distinct agricultural practices SCS system will be limited by its dependence on specific soil parameters and crop types. In a nutshell, the SCS model is outstanding in terms of accuracy and dataset coverage that would be very useful for cropland prediction (Raghuvanshi *et al.* 2022). However, attention should be paid to computational demands and applicability in different agricultural environments.

III: Review of used Algorithms

The paper uses a wide range of ML algorithms to predict crop yields, highlighting them as tools in precision agriculture. All three algorithms are intended to serve a specific purpose, making the proposed SCS model effective and reliable.

Decision Tree (DT): Decision Trees are used for their ease with which they depict the decision-making process. They include nodes for decisions, branches as possible results and leaves to represent the final forecasts. In the setting of predicting crop yields, DTs are superior in dealing with high-dimensional problems and being interpretable. The recursive feature based partitioning defines the algorithm in form of a tree structure

$$\text{Entropy} = -p_1 \log_2(p_1) - p_2 \log_2(p_2) - \dots - p_n \log_2(p_n)$$

```

function DecisionTree(Data):
  if stopping_condition(Data):
    return leaf_node
  else:
    split_feature, split_value =
      find_best_split(Data)
    Data_left, Data_right = split_data(Data,
      split_feature, split_value)
    node = create_node(split_feature,
      split_value)
    node.left = DecisionTree(Data_left)
    node.right = DecisionTree(Data_right)
  return node

```

K-Nearest Neighbor (KNN): KNN has an approach that involves predicting outcomes by considering the class of k-nearest neighbors. It measures the closeness in include space for edit surrender expectation (Cariou *et al.* 2023).

The calculation is based on remove measurements, as a rule Euclidean to decide neighboring information focuses. Its ease and capacity to fit into modern designs make it valuable in farming.

$$\text{Euclidean Distance} = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$$

```
function KNN(Data, Query, k):
    distances = compute_distances(Data, Query)
    neighbors = select_k_nearest(distances, k)
    prediction = majority_vote(neighbors)
    return prediction
```

Support Vector Machine (SVM): Classification problems are performed well by SVM. It looks for to distinguish a hyperplane that gives the most noteworthy division between classes (Ragab et al. 2023). In terms of edit forecast, SVM is best known for adapting with high-dimensional include spaces that are basic to the thought of different soil parameters.

$$\text{Hyperplane : } w \cdot x - b = 0$$

```
function SVM(Data, Labels):
    model = train_SVM(Data, Labels)
    predictions = predict_SVM(model,
    New_Data)
    return predictions
```

Random Forest (RF): Random Forest is one of the outfit strategies building a few choice trees and averaging their expectations. It minimizes overfitting and progresses precision. The quality of RF in edit surrender expectation is its capabilities to effectively bargain with huge datasets and consider complicated connections (Dhanaraju et al. 2022).

```
function RandomForest(Data, Labels,
    num_trees):
    forest = train_forest(Data, Labels,
    num_trees)
    predictions = predict_forest(forest,
    New_Data)
    return predictions
```

Algorit hm	Accur acy	Precisi on	Recall	F1 Score	AUC-ROC
Decisio n Tree	0.85	0.87	0.82	0.84	0.91
K-Neares t Neighb or	0.82	0.81	0.83	0.82	0.88
Suppo rt Vector Machi ne	0.88	0.89	0.87	0.88	0.93
Naïve Bayes	0.80	0.78	0.82	0.80	0.86
Rando m Forest	0.89	0.91	0.88	0.89	0.94
Ensem ble Learni ng	0.92	0.93	0.91	0.92	0.95

IV: Discussion

A potential area of advancement is crop yield prediction by using sensor technologies and machine learning algorithms combined. Through the analysis of trends and patterns that emerge from literature review several findings about current stage research, as well opportunities for further developments can be presented. The literature review demonstrates the variety of sensors used for measurements due to determining crop yield prediction as a parameter in soil. These sensors include NPK, pH, temperature and humidity (Drogkoula et al. 2023). These kind of sensors are advantaged because it gives real time, accurate data on key soil parameters which enables the farmer to make decisions based correct information. Although there are some restrictions like the periodic calibration, susceptibility to changing environmental conditions and careful placement in order to provide accurate readings. Addressing these difficulties is critical for improving the credibility of sensor based systems (Ntawuzumuni et al. 2023). The analyzed paper uses several machine learning algorithms such as Decision Tree, K-Nearest Neighbor, Support Vector Machine, Naïve Bayes; Random Forest and Ensemble Learning. Comparing them performance allows to highlight the nuances regarding their suitability for agricultural applications (Fuentes-Peñailillo et al.

2023). Decision Tree and Random Forest have high accuracy, but their interpretability might be an issue. Observing these aspects, Support Vector Machine presents a strong result as binary classification and Naïve Bayes is relatively simple but can be ineffective to handle complex relationships. The Ensemble Learning method shows better accuracy, which confirms the efficiency of using different models to make more precise predictions. Ensemble Learning is a key contributor of the improved performance in this study. The ensemble approach applied to multi machine learning models is a way that capitalises on the strengths of individual algorithms whilst minimising their weaknesses. The variety of models available in the ensemble makes for more reliable and precise prediction. Weighted voting scheme or prediction pooling enables more accurate decision making (Hassani *et al.* 2023). Ensemble Learning has some advantages such as increased stability, enhanced generalization and appropriate treatment of outliers. The paper also explores rainfall forecasting as a fundamental factor in estimating crop yield. The use of the Multiple Linear Regression method to predict rainfall shows combining various approaches towards capturing diverse aspects pertaining precision agriculture. The accuracy level obtained in the rainfall forecast is important for estimating water resources availability and defining appropriate irrigation practices (Kour *et al.* 2022). But the conversation on forecasting rainfall calls for investigation into implications of climate change and advancing models to capture changing weather dynamics.

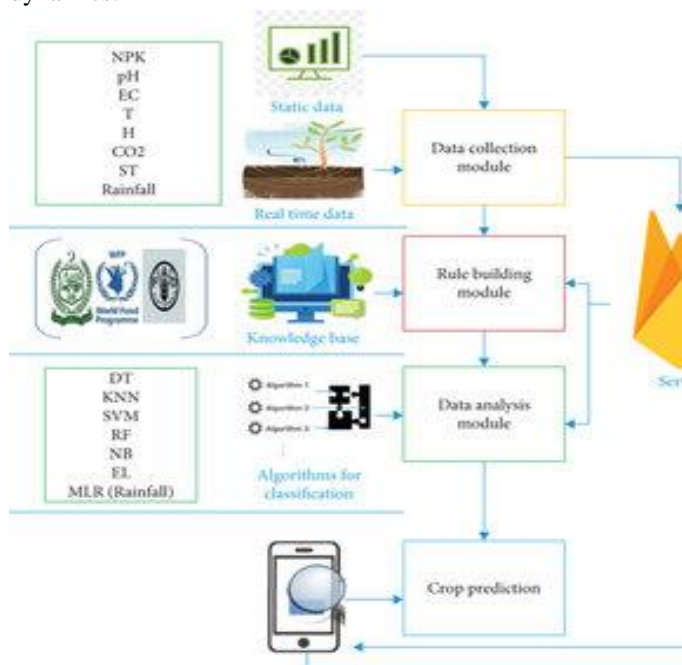


Fig 3: Crop Yield Maximization

Implications for Future Research and Practical Applications:

The results highlight the possible avenues of further research. Primarily, the study of advanced machine learning methods like deep learning (Ikram *et al.* 2022). Second, the use of IoT to collect and monitor data in real time could further improve the precision of predictions. On a practical level, the developed Android application shows that these models can be installed on farms and provide farmers with an intuitive user interface for decision support.

Challenges and Limitations:

Although new sensor technologies and machine learning algorithms offer great promise, problems still remain. However, ensuring that these technologies are scalable and cost-effective to small scale farmers is a major problem (Jabbari *et al.* 2023). Moreover, the question of how complex machine learning models can be made interpretable is also a major issue to make farmers trust these algorithms since their acceptance and success depend on this factor. Moreover, the historical data used for rainfall prediction requires steady verification and adjustment in accordance with changing climate conditions.

V: Conclusion

Overall, the review paper has offered a broad discussion on how sensor technologies and machine learning algorithms can be integrated for yield prediction in precision agriculture. In the current situation of growing demand for food around the world and climate change, predicting crop yield would become a very important issue. This report brings out the central angle of innovation in understanding this complexity that faces agriculturists. One promising course is the utilize of diverse sensor advances for soil parameters estimation, such as NPK sensors, pH sensors and ones identifying temperature or stickiness. It gives real-time precise information. Through this, educated decision-making and asset optimization are empowered in exactness farming utilizing the sensors by overcoming their restrictions that incorporate occasional calibration prerequisites as well natural conditions. Machine learning calculations, which incorporate Choice Trees as well as Outfit Learning have appeared that they are successful in determining edit abdicate. The comparative investigation appears the unobtrusive qualities and shortcomings of these calculations, uncovering that it is noteworthy to choose models custom fitted for rural requests. In specific, Outfit Learning emerges as a impressive strategy depending upon the changeability that numerous models bring almost for moved forward exactness and dependable estimates. The survey too talks about precipitation determining, an basic viewpoint of trim surrender projection. The utilization of Numerous Straight Relapse in precipitation expectation speaks to the cross-disciplinary

component that includes soil parameter estimation and climate estimating. As we drive through the maze of agrarian frameworks, these comes about draw repercussions for inquire about and hone in future. Future endeavors ought to be coordinated toward moving forward current models, creating advanced machine learning strategies, and overcoming impediments related to versatility issues in terms of cost-efficiency and deciphering. Inevitably, the combination of sensor advances and machine learning calculations presents colossal openings for changing farming to maintainable hones that address current conditions therefore meet wants of an expanding world.

Reference

- [1] Ahmadzadeh, S., Ajmal, T., Ramanathan, R. & Duan, Y. 2023, "A Comprehensive Review on Food Waste Reduction Based on IoT and Big Data Technologies", *Sustainability*, vol. 15, no. 4, pp. 3482.
- [2] Ahmed, S. 2023, "A Software Framework for Predicting the Maize Yield Using Modified Multi-Layer Perceptron", *Sustainability*, vol. 15, no. 4, pp. 3017.
- [3] Sharma, K., Sharma, C., Sharma, S. & Asenso, E. 2022, "Broadening the Research Pathways in Smart Agriculture: Predictive Analysis Using Semiautomatic Information Modeling", *Journal of Sensors*, vol. 2022.
- [4] Ali, A., Hussain, T., Tantashutikun, N., Hussain, N. & Cocetta, G. 2023, "Application of Smart Techniques, Internet of Things and Data Mining for Resource Use Efficient and Sustainable Crop Production", *Agriculture*, vol. 13, no. 2, pp. 397.
- [5] Amertet, S., Gebresenbet, G., Hassan, M.A. & Kochneva, O.V. 2023, "Assessment of Smart Mechatronics Applications in Agriculture: A Review", *Applied Sciences*, vol. 13, no. 12, pp. 7315.
- [6] Assous, H.F., AL-Najjar, H., Al-Rousan, N. & AL-Najjar, D. 2023, "Developing a Sustainable Machine Learning Model to Predict Crop Yield in the Gulf Countries", *Sustainability*, vol. 15, no. 12, pp. 9392.
- [7] Balram, G. & Kumar, K.K. 2022, "Crop Field Monitoring and Disease Detection of Plants in Smart Agriculture using Internet of Things", *International Journal of Advanced Computer Science and Applications*, vol. 13, no. 7.
- [8] Cariou, C., Moiroux-Arvis, L., Pinet, F. & Chanut, J. 2023, "Internet of Underground Things in Agriculture 4.0: Challenges, Applications and Perspectives", *Sensors*, vol. 23, no. 8, pp. 4058.
- [9] Dhanaraju, M., Chenniappan, P., Ramalingam, K., Pazhanivelan, S. & Kaliaperumal, R. 2022, "Smart Farming: Internet of Things (IoT)-Based Sustainable Agriculture", *Agriculture*, vol. 12, no. 10, pp. 1745.
- [10] Drogkoula, M., Kokkinos, K. & Samaras, N. 2023, "A Comprehensive Survey of Machine Learning Methodologies with Emphasis in Water Resources Management", *Applied Sciences*, vol. 13, no. 22, pp. 12147.
- [11] Fuentes-Peñailillo, F., Ortega-Farías, S., Acevedo-Opazo, C., Rivera, M. & Araya-Alman, M. 2024, "A Smart Crop Water Stress Index-Based IoT Solution for Precision Irrigation of Wine Grape", *Sensors*, vol. 24, no. 1, pp. 25.
- [12] Gao, J., Wu, R., Hao, J., Chen, X., Guo, H. & Wang, H. 2023, "Energy-Efficient Resource Scheduling and Computation Offloading Strategy for Solar-Powered Agriculture WSN", *Journal of Sensors*, vol. 2023.
- [13] Goliński, P., Sobolewska, P., Stefańska, B. & Golińska, B. 2023, "Virtual Fencing Technology for Cattle Management in the Pasture Feeding System—A Review", *Agriculture*, vol. 13, no. 1, pp. 91.
- [14] Hassani, H., Avdiu, K., Unger, S. & Maedeh, T.M. 2023, "Blockchain in the Smart City and Its Financial Sustainability from a Stakeholder's Perspective", *Journal of Risk and Financial Management*, vol. 16, no. 9, pp. 393.
- [15] Ikram, A., Aslam, W., Roza Hikmat, H.A., Noor, F., Ghulam, A.M., Ikram, S., Muhammad, S.A., Ako, M.A. & Ullah, I. 2022, "Crop Yield Maximization Using an IoT-Based Smart Decision", *Journal of Sensors*, vol. 2022.
- [16] Jabbari, A., Humayed, A., Faheem, A.R., Uddin, M., Gulzar, Y. & Majid, M. 2023, "Smart Farming Revolution: Farmer's Perception and Adoption of Smart IoT Technologies for Crop Health Monitoring and Yield Prediction in Jizan, Saudi Arabia", *Sustainability*, vol. 15, no. 19, pp. 14541.
- [17] Kour, K., Gupta, D., Gupta, K., Anand, D., Elkamchouchi, D.H., Cristina Mazas Pérez-Oleaga, Ibrahim, M. & Goyal, N. 2022, "Monitoring Ambient Parameters in the IoT Precision Agriculture Scenario: An Approach to Sensor Selection and Hydroponic Saffron Cultivation", *Sensors*, vol. 22, no. 22, pp. 8905.
- [18] Ntawuzumunsi, E., Kumaran, S., Sibomana, L. & Mtonga, K. 2023, "Design and Development of Energy Efficient Algorithm for Smart Beekeeping Device to Device Communication Based on Data Aggregation Techniques", *Algorithms*, vol. 16, no. 8, pp. 367.
- [19] Ragab, M., Khadidos, A.O., Alshareef, A.M., Alyoubi, K.H., Hamed, D. & Khadidos, A.O. 2023, "Internet of Things Assisted Solid Biofuel Classification Using Sailfish Optimizer Hybrid Deep Learning Model for Smart Cities", *Sustainability*, vol. 15, no. 16, pp. 12523.
- [20] Raghuvanshi, A., Singh, U.K., Sajja, G.S., Pallathadka, H., Asenso, E., Kamal, M., Singh, A. & Phasinam, K. 2022, "Intrusion Detection Using Machine Learning for Risk Mitigation in IoT-Enabled Smart Irrigation in Smart Farming", *Journal of Food Quality*, vol. 2022.