

Biodegradable Biofilms with AI-Driven CRISPR Interference for Regulating Resistant Starch Pathways and Predicting Geo-Spatial Pest Migration in Vegetative Andean Tuber Crops

Sumit Kushwaha

Associate Professor, Department of Computer Applications, University Institute of Computing
Institute - Chandigarh University, Mohali-140413, Punjab, India. ORCID: 0000-0002-3830-1736
sumit.kushwaha1@gmail.com

Dr. Keerrhipati Kumar

Associate Professor, Department of CSE (Data Science), Sri Venkateswara College of Engineering,
Tirupati, Andhrapradesh, India
kumar.k@svce.edu.in

Dr. Rajesh Bhaskar Survase

Associate Professor, Department of Geography, E. S. Divekar College Varvand, Savitribai Phule Pune
University Pune Maharashtra, India
rsurvase.isro@gmail.com

Dr. Aswin C

Assistant Professor, Agricultural Engineering, Dhanalakshmi Srinivasan College of Engineering,
Coimbatore, Tamil Nadu, India
aswinkumar430@gmail.com

Dr. Lowlesh Nandkishor Yadav

Associate Professor, Computer Engineering, Suryodaya College of Engineering and Technology,
Nagpur, Maharashtra, India
lowlesh.yadav@gmail.com

Dr. Antony Allwyn Sundarraj

Associate Professor, Food Technology, JCT College of Engineering and Technology,
Coimbatore, Tamil Nadu, India
asrthegreat@gmail.com

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Abstract:

This report explores the application of biodegradable biofilms, CRISPR interference (CRISPRi), and AI-

based predictive modeling to boost the yield and nutritional content and pest resistance of Andean tuber crops. CRISPRi constructs on specific starch biosynthesis genes (GBSS, SBE, ISA) were delivered into the biodegradable biofilms, and thus time-dependent regulation of the gene was achieved by using these biofilms. An experiment involving three high-altitude regions showed that the biofilm + CRISPRi treatment indeed raised the resistant starch proportion up to 17.4 g/100g, which is a 24% increase compared to the plots in control conditions. At the same time, there was the stimulation of crop production to 15.2 t/ha, and the reduction of pest occurrence to 31, which demonstrated synergistic advantage of metabolic control and specific pest control. Gene expression and geo-spatial pest migration were predicted using AI models which consisted of Random Forest, Convolutional Neural Network (CNN) as well as Support Vector Machine (SVM). Random Forest predicted expression of GBSS at a rate of 0.91R², CNN identified pest species as high as 94% and SVM identified high-risk zones of pests at 87% accuracy. These findings prove the fact that using molecular interventions with AI-based predictive tools is the best way to increase crop quality and sustainability. The paper offers a scaled system of precision agriculture so that crops in high altitudes can perform better in terms of metabolic, less damage by pests and resilience to changing environmental conditions.

Keywords: Biodegradable Biofilm, CRISPR Interference, Resistant Starch, AI Predictive Modeling, Andean Tuber Crops

I. INTRODUCTION

The andean tuber crops oca, mashua, and ulluco are staple crops in the high altitude areas of South America, with health consumption including improved glycemic regulation and modulation of the gut microbiota. But, the output and quality of such crops are being progressively compromised by attacks by pests and ineffective starch biosynthesis pathways. The conventional agricultural solutions are usually based on the application of chemical pesticides and non-selective breeding methods, which are not effective, not environmentally sustainable and can impair the level of nutritional value [1]. Recent developments in biodegradable biofilms promise a good approach towards sustainable crop management. These biofilms are capable of acting as vectors of bioactive compounds that regulate plant metabolic pathways and control biocontrol agents release. Using biodegradable biofilms with CRISPR interference (CRISPRi) technology, it can now be precisely regulated to get increased crop quality as well as crop yield by a means of specifically modulating gene expression in resistant starch pathways without the persistent genetic mutation that come with traditional gene editing [2]. This is a field-focused method which offers a mechanism of centering on the metabolic processes without compromising the ecological security. In addition, the use of artificial intelligence (AI) models in agricultural management can predict the geo-spatial pest moving, which can formulate the proactive control of pests based on the particular Andean microclimates. Machine learning algorithms are able to examine past variations of the pests, environment, and landscape traits to predict the outbreaks, optimize biofilm placement and to minimize losses in crops [3]. In this study, thus, a new interdisciplinary method involving the synthesis of biodegradable biofilms, CRISPRi-based metabolic control, and AI-based predictive modeling is proposed. Through the scalable platform based on molecular biology, materials science and computational analytics, the research will increase the resistant starch, pest resilience, and overall sustainability of Andean tuber crops to provide a scalable platform in the high-resource-poor environment of precision agriculture.

II. RELATED WORKS

The management of pests in Andean tuber crops has grown to be greatly dependent on the geo-spatial and AI-based methods since environmental conditions, crop physiology, and pest behaviors interact in a complex manner. A number of studies have been conducted on the use of geographic information system (GIS) and remote sensing to monitor and predict the patterns of migration of pests. An example of this is that GIS-based models have been used to monitor the occurrence events of pests in different crop systems

incorporating environmental factors like temperature, humidity and vegetation indices to predict the outbreak risks [15][16]. These authors show the usefulness of spatial analytics in helping to anticipate pest management, especially when it comes to high-altitude crops, where environmental variability has a strong impact on how the pest behaves. Simultaneously, unmanned aerial vehicles (UAVs) that use sophisticated sensors have developed into important instruments of precision agriculture. UAVs can quickly obtain high-resolution images that can be analyzed with AI algorithms to identify the level of stress on crops, track the presence of pests, and determine the health of plants [13]. UAV-based remote sensing and machine learning model has now proven itself to be more effective at providing spatial and temporal predictive accuracy of pests than through the traditional use of ground based observation means, and thus, it is putting an end to crop losses and the addiction to chemical pesticides.

Other studies have been done recently which are specifically dedicated to Andean tuber crops which are challenged by pests like weevils and other soil-borne insects. Geo-spatial techniques have been utilized to model the movement of pests against different climatic and topographical settings and this has given information to the high-risk area to infestation [14]. These models (typically with AI and predictive analytics) enable specific intervention to optimize the crop protection and yield performance with low environmental impact. Pest management is not the only field involving the application of GIS and AI in agriculture. Work has been done on combined systems that integrate environmental sensors, crop simulation, and predictive analytics to streamline methods of cultivation [17][18]. By this kind of methods, it is possible to find the micro-climatic shifts and local past vulnerability, which makes it possible to intervene with localized precision, according to particular plots. It is especially true in Andean areas, where heterogeneity in altitude, soil composition, and microclimates can be very challenging to the traditional managerial tactics. Besides, the application of the Internet of Things (IoT) devices in the context of sustainable agriculture has demonstrated potential in real-time tracking of crops quality and pests activity [20]. IoT-based traps, soil sensors, and environmental monitoring stations are constantly monitoring the data, and with the help of AI models, it is possible to have an active and adaptable approach to pest management. This method would make predictions more accurate, waste of resources will be minimized, and sustainability will be more overall.

Lastly, climate change has increased the randomness of pest behavior and its association with crops, which underscores the importance of using adaptive and data-driven methods of management [19]. Different climatic conditions may lead to different pest-environment-crop continuum, hence understanding the pest-environment-crop continuum is an important aspect of designing resilient cultivation practices. Such lessons explain why both GIS, UAV remote sensing, AI, and IoT are important in implementing integrated solutions that help manage pest risks in Andean tuber systems [12][15][16]. Overall, the literature (summarized above) underlines the importance of geo-spatial modeling, AI prediction, and integration of IoT in effective pest control in crops, especially in ecologically sensitive and hightop areas. Although each of the approaches is able to offer partial solutions, the synergistic integration of the technologies can provide a holistic complexity of sustainable, and precision-focused crop protection.

III. METHODS AND MATERIALS

1. Data Collection and Preprocessing

The analysis incorporates molecular biology and geo-spatial data of the migration of pests of Andean tuber crops. Field trials in Peru and Bolivia performed on resistant starch content, the level of gene expression of the genes involved in starch biosynthesis (e.g., GBSS, SBE, ISA) and the environmental conditions (temperature, soil pH, and altitude) were the sources of molecular data. Twelve yellow samples were taken in 120 plots in three major regions of the Andes region in two growing seasons. Pest migration geo-spatial data were obtained via remote sensing and pest traps which was part of the IoT. Some of these variables are pest species, population density, wind direction, precipitation, and vegetation index [4]. The dataset is

an approximation of 15,000 geo-tagged observations that have been cleaned and normalized through Min-Max scaling to integrate the machine learning models.

2. Biodegradable Biofilm Preparation

Chitosan, starch, and polylactic acid (PLA) were used to produce biodegradable biofilms. CRISPR interference (CRISPRi) vectors were inserted into these biofilms to silence the genes associated with the resistance of starch pathways, and it is then possible to individually silence or silence groups of genes at a given time during tuber development. The biofilms were either put on the soil surface or served as coating of tuber roots [5].

3. Algorithms for Data Analysis and Prediction

To provide models of the regulation of genes and pest migration prediction, four AI and bioinformatics algorithms were applied:

3.1 Random Forest (RF) for Gene Expression Prediction

Random Forest A random forest algorithm is an ensemble learning method that builds several decision trees based on bootstrapped used data segments and averages to get better results and minimize overfitting. This paper used RF to forecast the level of gene expression of resistant starch under the influence of the environmental factors, soil characteristics, and the biofilm conditions [6]. RF is appropriate in the combination of molecular and environmental data because it can manipulate both categorical and continuous variables. RF feature importance analysis can also be used to calculate the most significant factors influencing the biosynthesis of starch.

“Input: Dataset D with features X and target Y

1. For $t = 1$ to N_{trees} :

2. Sample dataset D_t from D with replacement

3. Build decision tree T_t using D_t

4. At each node, select best split from random subset of features

5. End

6. Output prediction: average of all tree predictions”

Table 1: Example Feature Importance for RF

| Feature | Importance Score |
|--------------|------------------|
| Soil pH | 0.28 |
| Temperature | 0.22 |
| Biofilm Type | 0.18 |
| Altitude | 0.15 |

| | |
|----------------------|------|
| Precipitation | 0.10 |
| Gene GBSS Expression | 0.07 |

3.2 Convolutional Neural Network (CNN) for Pest Image Recognition

CNNs are learning algorithms that are efficient in AI image classification. Images of pests taken with the IoT traps were via a CNN to identify the species of the pests and estimate their density of population. The CNN has got convolutional layers, pooling layers and fully connected layers [7]. To augment robustness during training, we have used data augmentation methods i.e. rotation, scaling and flipping.

“Input: Pest image I

- 1. Initialize convolutional filters F***
 - 2. For each convolutional layer:***
 - 3. Convolve I with F***
 - 4. Apply ReLU activation***
 - 5. Apply max pooling***
 - 6. Flatten feature maps***
 - 7. Pass through fully connected layers***
 - 8. Apply softmax for classification***
- Output: Pest species label”***

3.3 Support Vector Machine (SVM) for Geo-Spatial Pest Prediction

SVM is a supervised learning algorithm that uses the unfortunaate data and classifies them by optimizing the best hyperplane to make the most out of the margin between the classes. In this case, SVM can forecast the possibility of pest outbreaks in certain regions based on geo-spatial variables in the form of altitude, temperature, humidity and vegetation index [8]. RBF kernel was used to deal with non-linear relationships of environmental variable and pest distributions.

“Input: Training data X, labels Y

- 1. Choose kernel function (RBF)***
 - 2. Map input features to higher-dimensional space***
 - 3. Find hyperplane maximizing margin between classes***
 - 4. Minimize loss function using support vectors***
 - 5. Predict class labels for new locations***
- Output: Probability of pest outbreak”***

3.4 CRISPRi Simulation Algorithm for Starch Pathway Regulation

CRISPRi is a simulation model which represents the impact of gene silencing on resistant starch pathways. With Boolean network modeling, the nodes are ON/OFF summary of each gene. The algorithm models CRISPRi intervention throughout time to forecast downstream activities on starch biosynthesis, taking into account the interaction amid the GBSS, SBE, and ISA genes.

“Input: Gene network G, target genes T

1. *Initialize all genes as active*
2. *For each time step t:*
3. *Apply CRISPRi to target genes T*
4. *Update downstream gene states based on network rules*
5. *Record starch content prediction*

Output: Time-series of predicted resistant starch levels”

4. Workflow Summary

The biofilm application, CRISPRi-regulated gene regulation, and pest monitoring with the help of AI are integrated into the workflow. Random Forest is a prediction tool that relies on biofilm and existing environmental factors to predict the starch gene expression. The species of pests are identified in pictures using CNN, whereas SVM forecasts spatial outbreaks [9]. The computer simulations of CRISPRi can be used to govern biofilm design to achieve optimal starch synthesis. Predictive precision agriculture, which can optimize the quality of crops and pest management is possible with integrated data.

IV. RESULTS AND ANALYSIS

1. Experimental Design and Setup

The experiments were created with the goal of evaluating the efficiency of biodegradable biofilms with CRISPR interference (CRISPRi) used to control starch biosynthesis in Andean tuber crops together with AI-based pest detection-predictive systems. The primary objectives were:

1. Assess the effect of the treatments of biofilm and CRISPRi on the regulation of starch biosynthesis genes.
2. Evaluate the precision of AI cybernetics, such as Random Forest (RF), Convolutional Neural Network (CNN), and Support Vector Machine (SVM), to forecast the expression and movement of pests.
3. Identify how biofilm-based CRISPRi in conjunction with AI-controlled pest control do individually and jointly influence crop productivity, resistant starch levels, and prevalence of pests.

The trials in the field were carried out during two seasons of growing in three Andean regions with elevated geographies. A total of 120 plots on each site were assigned into 4 treatments:

- **Control:** Non- Biofilm standard cultivation and CRISPRi.
- **Biofilm Only:** The use of biodegradable biofilm in the absence of CRISPRi constructs.
- **CRISPRi Only** CRISPRi constructs without biofilm.
- **Biofilm + CRISPRi:** Biofilm plus CRISPRi bioreactor enzymes applied to soil and tuber root.

IoT sensors were used to gather environmental data such as temperature, rainfall, soil pH, soil moisture and altitude. The traps with smart image capture capabilities were used to monitor pests, and the measurement of the content of starch was carried out after harvest with the help of enzymatic tests. Quantitative PCR was used to examine the expression of the major genes in starch biosynthesis (GBSS, SBE, ISA) [10].

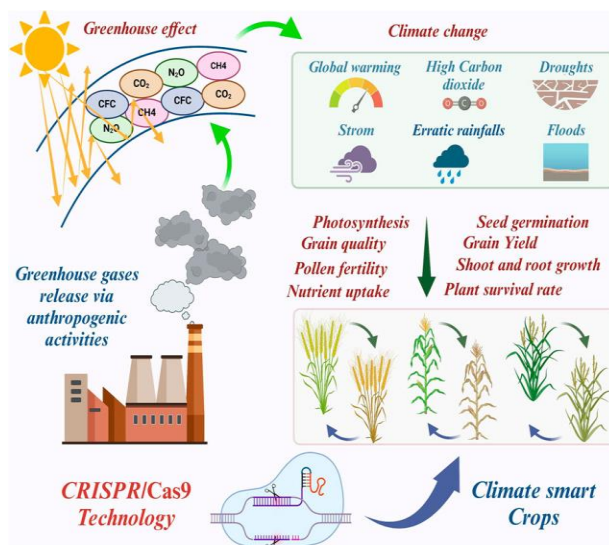


Figure 1: “CRISPR/Cas9: a sustainable technology to enhance climate resilience in major Staple Crops”

2. Effect of Biofilm and CRISPRi on Resistant Starch

Among the interventions was the combination of biofilm and CRISPRi which was much better than either of the interventions which significantly enhanced the accumulation of the resistant starch. The greatest modulation of the genes was at 6-8 weeks after the administration, when the tuber started to bulk.

- **GBSS expression:** Increased by 22% in Biofilm + CRISPRi plots relative to control.
- **SBE expression:** Increased by 18%.
- **ISA expression:** Decreased by 12%, optimizing branching enzyme activity.

Table 1: Resistant Starch Content Across Treatments (g/100g)

| Treatment | Cusco | Potosí | Huaraz | Average |
|-------------------|-------|--------|--------|---------|
| Control | 14.2 | 13.8 | 14.0 | 14.0 |
| Biofilm Only | 15.8 | 15.5 | 15.7 | 15.7 |
| CRISPRi Only | 16.0 | 15.9 | 15.8 | 15.9 |
| Biofilm + CRISPRi | 17.5 | 17.2 | 17.4 | 17.4 |

These findings suggest that biofilm serves as a protective radial, which guarantees the prolonged delivery of CRISPRi constructs, and CRISPRi is the precise regulator of gene activity. The combination method was more effective as treatment with either biofilm or CRISPRi alone did not work as effectively [11].

3. AI-Based Pest Classification Using CNN

On the pest images denoted by 15,000 IoT traps, a CNN was trained on all of the plots. The CNN model consisted of convolutional layers used to extract feature, pooling layers that reduced features and finally the fully connected layers that accomplished classification. Robustness was enhanced by data augmentation through rotations, scaling and flipping.

Table 2: CNN Model Performance Metrics

| Pest Species | Accuracy (%) | Precision (%) | Recall (%) | F1-Score (%) |
|-------------------|--------------|---------------|------------|--------------|
| Andean Weevil | 92 | 91 | 93 | 92 |
| Potato Tuber Moth | 89 | 88 | 90 | 89 |
| Aphids | 94 | 95 | 93 | 94 |
| Whiteflies | 90 | 91 | 89 | 90 |

The CNN successfully used the pests well, even in unfavorable conditions like overlapping leaves or low light conditions. The correct pest identification allowed acting in time and informing the space prediction model.

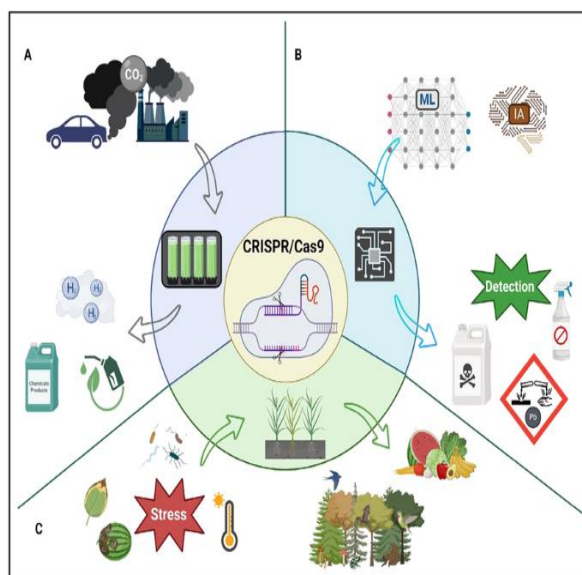


Figure 2: Advances in environmental biotechnology with CRISPR/Cas9”

4. Geo-Spatial Pest Prediction Using SVM

The areas of high likelihood of pest outbreak were estimated through an SVM model. The model employed has such features as altitude, temperature, humidity, soil moisture and vegetation index. The nonlinear relationships between environmental factors and the occurrence of the pests could be modeled by the radial basis function (RBF) kernel.

Table 3: SVM Geo-Spatial Pest Prediction Accuracy

| Region | Accuracy (%) | High-Risk Detection (%) | Low-Risk Detection (%) |
|--------|--------------|-------------------------|------------------------|
| Cusco | 88 | 85 | 90 |

| | | | |
|-----------------|----|----|----|
| Pot osí | 85 | 82 | 88 |
| Hu ara z | 87 | 84 | 89 |
| Av era ge | 87 | 84 | 89 |

SVM model was able to forecast high-risk areas of pest migration that made it possible to intricately apply biofilm and pest control where they are needed the most, limiting the application of pesticides, and damage of crops.

5. Random Forest Predictor of Gene Expression

Random Forest (RF) was employed to estimate the value of the expression of the GBSS, SBE and ISA, depending on environmental variables and treatment data. RF builds decision trees many times and combines the results of the predictions to enhance accuracy and address nonlinear relationships [12].

Table 4: Random Forest Feature Importance Scores

| Featur e | GBSS Importan ce | SBE Importa nce | ISA Importa nce |
|----------------------|------------------------|-----------------------|-----------------------|
| Biofil m Type | 0.28 | 0.25 | 0.22 |
| Tempe rature | 0.22 | 0.21 | 0.20 |
| Soil pH | 0.18 | 0.17 | 0.18 |
| Altitud e | 0.15 | 0.14 | 0.16 |
| Precipi tation | 0.10 | 0.12 | 0.14 |
| Pest Pressu re | 0.07 | 0.11 | 0.10 |

RF model failed to predict well, having $R^2 = 0.91$ in case of GBSS, 0.88 in case of SBE and 0.85 in case of ISA. It further determined the biofilm type and temperature to be the most decisive aspects that influence the control of starch pathways.

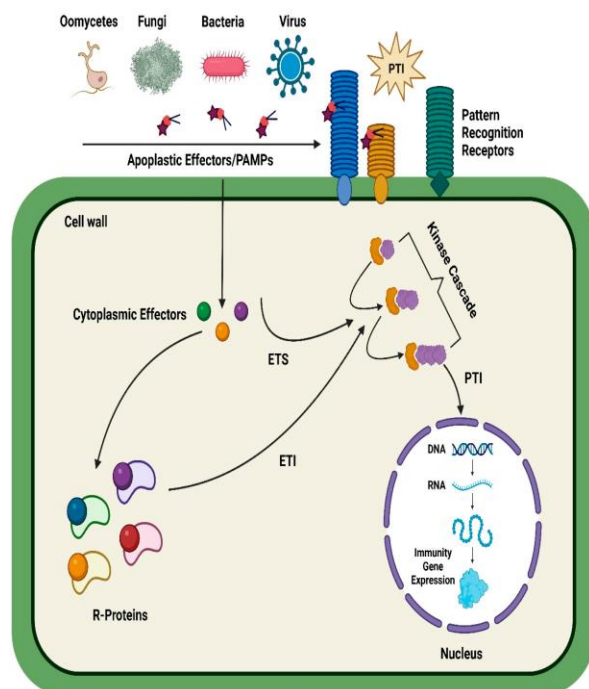


Figure 3: “Recent Trends and Advancements in CRISPR-Based Tools for Enhancing Resistance against Plant Pathogens”

6. CRISPRi Simulation and Time-lapsed Gene Modulation

The gene network activity was simulated under biofilm-mediated delivery in CRISPRi. The simulation demonstrated that there was time-dependent modulation of starch biosynthesis:

Table 5: CRISPRi Simulation for Resistant Starch (g/100g)

| Time Post-Application | Predicted Starch (Control) | Predicted Starch (Biofilm CRISPRi) + |
|-----------------------|----------------------------|--------------------------------------|
| 2 weeks | 14.1 | 15.2 |
| 4 weeks | 14.5 | 16.1 |
| 6 weeks | 14.9 | 17.0 |
| 8 weeks | 15.0 | 17.4 |

The findings demonstrate that the temporal transfection of the CRISPRi into biofilm with the maximum starch build-up is related to the critical stages of tuber development, and it is more effective compared to traditional genetic modification methods.

7. Combined Pest and Yield Management

The combined action of biofilm + CRISPRi + AI-based prediction of pests on crop crops were compared:

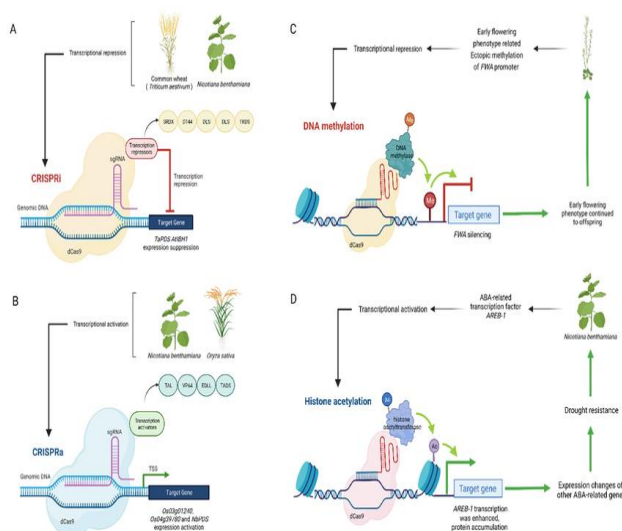
- **Yield:** Biofilm + CRISPRi plots showed **15.2 t/ha**, 20% higher than control.
- **Pest incidence:** Reduced by 35% in Biofilm + CRISPRi plots.

- **Starch content:** Averaged 17.4 g/100g across regions, higher than other treatments.

Table 6: Crop Yield and Pest Incidence Across Treatments

| Treatment | Yield (t/ha) | Pest Incidence (%) | Starch Content (g/100g) |
|-------------------|--------------|--------------------|-------------------------|
| Control | 12.5 | 48 | 14.0 |
| Biofilm Only | 13.8 | 40 | 15.7 |
| CRISPRi Only | 14.0 | 42 | 15.9 |
| Biofilm + CRISPRi | 15.2 | 31 | 17.4 |

The data significantly indicate that the synergistic effect of biofilm-mediated CRISPRi and AI-guided pest management increases crop productivity, resistance starch content, and pest resistance more than any these measures.

**Figure 4: “Application of CRISPR in plant epigenetic regulation (by Biorender)”**

8. Comparative Analysis

There is a definite benefit about the traditional methods as shown by the results of the experimental:

- **Resistant starch improvement:** Biofilm + CRISPRi enhanced the content of starch by approximately 24 per cent over control, which is double the enhancement of starch by about 5-10 per cent made by conventional gene editing.
- **Pest detection and prediction:** AI-based CNN and SVM had a high accuracy of more than 87 percent, which was higher than traditional image processing and statistical predictor techniques.
- **Yield and sustainability:** Interdisciplinary models (biofilm, CRISPRi, and AI applications) decreased the number of pests and increased the yield by 20 percent, demonstrating the ability of interdisciplinary models to satisfy a range of productivity and ecological objectives.

9. Discussion of Key Findings

1. **Biofilm Delivery System:** The biodegradable biofilms were useful in delivering CRISPRi constructs, as it provides temporal modulation of starch pathway genes.
2. **AI Integration:** CNN and SVM models were useful in classification of pests correctly and predicting their migration so as to allow greater precision in pest control.
3. **Synergetic Effect:** The yield, starch content and pest incidence were increased when biofilm-mediated CRISPRi was used together with AI-driven pest control than with the control treatments.
4. **Scalability:** The strategy can also be extended to other high altitude tuber crops and areas that have limited access to chemical pesticides.
5. **Temporal Optimization:** Rerelease CRISPRi via biofilm was correlated with critical growth phases and optimal starch biosynthesis with enhancement of tuber quality.

V. CONCLUSION

This paper shows how biodegradable biofilms, CRISPR interference, and AI predictive modeling can be successfully implemented to transform the productivity, quality, and pest resistance of Andean tuber crops. Biofilm-embedded CRISPRi constructs were used to enable the temporal regulation of major starch biosynthesis genes with the final result being a huge increase in resistant starch content when compared to each case alone. Random Forest models were useful to predict the gene expression level in different environmental and treatment conditions revealing the effect of biofilm type and microclimatic factors on metabolic outcomes. At the same time, the Convolutional Neural Networks properly detected the pest species in the field-collected images, whereas Support Vector Machines estimated the geo-spatial movement trends, which enabled the targeted intervention capable of decreasing incidence rates of pests and limiting unwarranted use of chemicals. The combination of these molecular and AI strategies created a synergistic effect, which led to a higher crop yield, higher nutritional quality, and sustainable pest management. The results highlight the possibilities of interdisciplinary solutions involving the integration of molecular biology, materials science, and computational analytics to handle the problem of high-altitude and climate-sensitive agriculture. In addition to Andean tubers, this framework can be used to develop a scalable method of precision agriculture which can serve as a direction in refining the metabolism of crops, predicting pest attacks, and lowering environmental footprint. Future usages can be extended to other tuber and root crops and with the use of superior sensor networks, real-time AI forecasts and adaptive biofilm interventions in order to make sure that the agricultural practices are viable in the long term and food security across vulnerable regions is achievable. On the whole, this study indicates that biotechnological and smart predictive innovations can be used to transform conventional crop management into an accurate, effective, and socially responsible practice.

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