



Drones, AI and Computer Vision: The Future of Precision Insect Pest Management

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A mong the various biological factors hindering sustainable agricultural production, insect pest problems rank high. The traditional methods of pest monitoring require field visits, visual tracking, and timely sampling, which are labour-intensive methods that may not provide timely information on wide agricultural regions. Modern technologies such as drones, artificial intelligence (AI), remote sensing, and computer vision are emerging technologies that offer opportunities for modern pest management of insects. The drones capture images of crops at high resolution while AI algorithms analyze these images to determine stress, hotspots, and infestation patterns. The computer vision models, including convolutional neural networks (CNN) and object detection systems, will recognize the insect species and their population. The information will allow the disbursement of pesticide at the specificity of the site as opposed to using the pesticide uniformly. As a result, the amount of pesticide used can be reduced simultaneously with the increasing scouting effectiveness, IPM efficacy, and decision-making. However, certain obstacles, such as problems associated with quality of the images, specificity of species identification, and preparation of datasets as well as model transferability, cost issues, legal regulations and field verification need to be tackled first. Precision entomology will be shaped by combining aerial observation with ground tracking, AI analysis and automated decision-making systems.

Keywords: Drones, Artificial Intelligence (AI), Remote Sensing, Integrated Pest Management (IPM)

Introduction

There is a potential for agricultural insect pests to contribute to sizeable yield losses. Effective pest management is therefore contingent on proper and timely detection, identification and intervention. Scouting is still significant in pest management, but obstacles such as the size of the fields, the manpower needed and variability of pest populations render continuous surveillance nearly impossible. Drones (also known as UAVs or Unmanned Aerial Vehicles) have emerged as an effective tool for gathering high-resolution data regarding the crop fields. Different drones equipped with an RGB, multispectral, hyperspectral or thermal sensors can transmit information that can indicate some changes in the colour of the crop or crop temperatures. The next step in the development of drones is

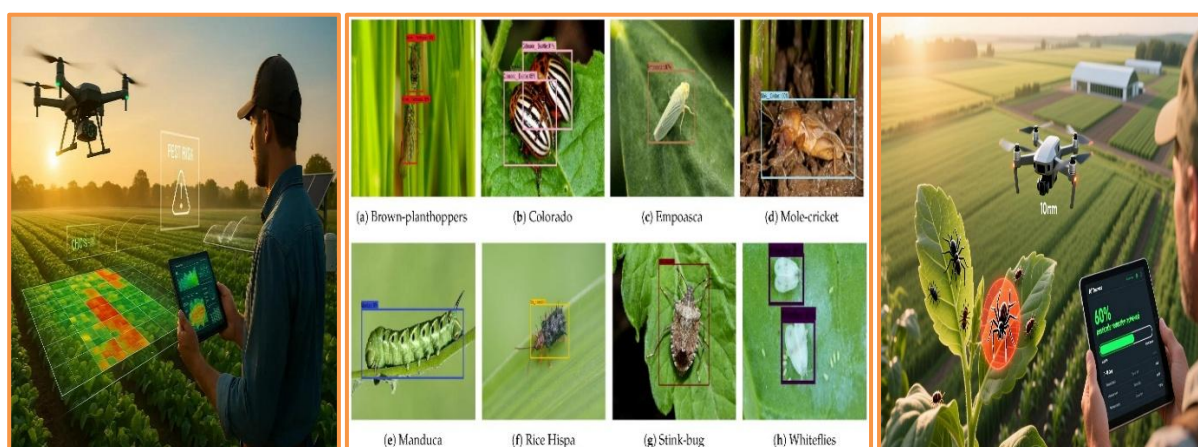
using AI and computer vision technology for the drones to go beyond mere image capturing and move to processing huge sets of images.

Interconnections Between Drones, Computer Vision and Artificial Intelligence

The process of precision pest management can be divided into and represented by the five steps:

Drone flight → Image acquisition → Computer vision → AI-based diagnosis → Site-specific management

The first step is performed when the drone is flying over the crops according to the given flight plan and making pictures. After that, the received images have to be processed to obtain the necessary information. Computer vision enables getting the information needed. Then, AI algorithms are applied to detect or classify pest insects, their patterns and areas of damage. The end product can be represented in the form of the digital pest risk map, which can be used by farmers/agricultural specialists to see if any further scouting or application of biological control measures or pesticides is required.



Role of Computer Vision in Pest Detection

Computer vision enables computers to interpret information contained in digital images. The detection, counting, and classification of insects can be achieved through the use of this technology in the field of entomology. For example, the images obtained from sticky traps can be analyzed and identified to automatically recognize the respective insect species. Similarly, close-up images of the leaves and canopies of crops may either show symptoms of damage caused by feeding or have the insect present in the captured images. Deep learning models such as CNNs have become important because they can learn visual characteristics directly from images. Object-detection approaches can identify the position of insects within an image, while classification models can assign an image to a pest category. Recent reviews indicate that CNN-based methods remain widely used, while newer approaches involving attention mechanisms and visual transformers are increasingly being investigated for fine-grained insect recognition.

Table 1. Major technologies and their role in precision pest management

Technology	Main function	Application in pest management
RGB camera	Captures visible images	Pest/damage recognition
Multispectral sensor	Captures selected spectral bands	Detection of crop stress
Hyperspectral sensor	Records detailed spectral information	Early stress and pest-related analysis
Thermal camera	Measures canopy temperature	Detection of physiological stress
GPS/GIS	Provides spatial coordinates	Pest hotspot mapping
Computer vision	Extracts visual features	Insect detection and counting

Technology	Main function	Application in pest management
Machine learning	Learns relationships in data	Pest classification/prediction
Deep learning	Advanced image analysis	Automated detection and recognition
IoT sensors	Provides field-level environmental data	Pest forecasting and decision support

From Pest Detection to Precision Management

The greatest advantage of AI-enabled pest monitoring is that detection can be linked with management decisions. A traditional method might entail spraying pesticides across an entire crop area in instances when pest occurrences are suspected. Nevertheless, a precision technique is intended to identify particular spots with considerable pest pressure. For instance, following drone imaging that shows areas with high risk in a corn plantation, personnel can examine the areas in question, thereby applying intervention methods only in places where economic thresholds have been surpassed. This creates a shift from “**spray first and inspect later**” to “**detect, verify and intervene precisely.**”

Conventional system

Field scouting → General pest observation → Whole-field decision → Uniform spraying

AI-enabled precision system

Drone survey → AI detection → Pest hotspot map → Ground verification → Threshold-based decision → Targeted intervention

Applications in Agricultural Entomology

The technology has several potential applications:

- Early Pest Detection
- Insect Counting
- Pest Hotspot Mapping
- Precision Spraying
- Pest Forecasting

Table 2. Comparison of conventional and AI-enabled pest monitoring

Parameter	Conventional monitoring	AI-enabled monitoring
Field coverage	Relatively limited	Potentially large
Data collection	Manual	Automated/semi-automated
Insect identification	Expert dependent	AI-assisted
Spatial information	Often limited	High spatial resolution
Monitoring frequency	Periodic	Potentially frequent
Pest hotspot detection	Difficult at large scale	Digital mapping possible
Decision support	Mainly observation-based	Data-driven
Labour requirement	Relatively high	Can be reduced
Scalability	Limited by manpower	Higher potential scalability

Advantages of Eco-Friendly Pest Control

The use of drones along with AI and computer vision can facilitate better practices in crop protection. Improved targeting may keep down the number of pesticide treatments made unnecessarily and the resulting environmental impact. Furthermore, agricultural staff can use the technology to allocate their field trips better and concentrate on specific problem areas.

Another significant advantage is that the technology can be integrated into IPM. AI should not supplant ecological knowledge or expertise, but complement it with data for deciding if and when intervention is warranted. Thus, precision pest management can lead to better efficiency in terms of resources used while continuing to protect crops.

Extensive Challenges

Despite fast advancements in technology, many downsides persist. To start with, the quality of images differs everywhere. Factors such as light conditions, presence of shadows, strong wind, density of standing crops and motion of insects may affect image recognition. Next, available labelling datasets are limited. It is crucial for algorithms to have datasets with images of various species of insects, stages of plant growth, backgrounds and nature conditions included. Thirdly, the transferability of field conditions stays a challenging point. A model that showed good results during one season may fail to present similar results at another time. Moreover, it is hard to catch small insects on the aerial pictures. In many cases, it is more efficient to use drones to define the degree of crop stress rather than to identify pests. Lastly, it may cost a lot to have such equipment and run it. In addition, trained professionals should operate the equipment.

Table 3. Challenges and possible solutions

Challenge	Possible solution
Poor-quality images	Better sensors and image preprocessing
Limited training data	Larger, diverse and field-based datasets
Similar-looking pest species	Fine-grained recognition models
Changing field conditions	Multi-season and multi-location validation
Small insect size	Ground cameras and trap-based imaging
High data volume	Edge computing and efficient AI models
High operational cost	Shared/custom service models
Model bias	Independent field testing
Regulatory limitations	Compliance with UAV regulations

Future Prospects

The future is likely to move from simple image-based detection toward integrated intelligent pest-management systems. Multisensor data fusion may combine RGB, multispectral, hyperspectral and thermal observations with weather and field sensors. AI models may also become more efficient and capable of running closer to the point of data collection. Autonomous drone systems and coordinated drone operations could further increase monitoring capacity. Another promising direction is the development of digital pest early-warning systems. It is possible to use drones in conjunction with the weather forecast, the crop growth stage, insect traps, and historical pest data to determine the likelihood of pest outbreaks. Future studies should focus more on independent field validation instead of solely depending on accuracy information obtained from the same datasets used in the model creation. Recent meta-analysis has highlighted substantial weaknesses in validation practices in UAV-based pest and disease studies, making robustness and transferability important priorities.

Conclusion

The utilization of drones, AI, and computer vision is revolutionizing precision insect pest management practices. Drones are capable of providing fast and accurate spatial information, in images, computer vision extracts any required details, while AI uses this knowledge to identify and classify pests. However, technology should not be perceived as an alternative to monitoring specialists, practitioners, and Integrated Pest Management methods but be treated instead as a supportive tool in their decision-making process. The future precision entomology will be determined by the availability of reliable datasets, effective AI models, reasonably priced sensing technologies, validation in the field, and proper integration with the integrated pest management system. When these components are combined, intelligent pest monitoring can move agriculture closer to a system in which pest problems are detected earlier and management actions are applied more selectively and efficiently.

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