

Intelligent Image-Based Defense Against Incursion from The Wild Animals

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Abstract

A revolutionary development in agricultural security, Smart Crop Guardian is designed expressly to counter invasions from wild animals. This system merges cutting-edge sensors and powerful algorithms to give an all-encompassing defense mechanism for crops by utilizing cutting-edge image-based technologies. Smart Crop Guardian continuously monitors the environment to detect and track wildlife hazards in real time, enabling prompt intervention and reducing possible harm. This abstract explores the salient characteristics and advantages of Smart Crop Guardian, showcasing its potential to transform crop protection techniques. The core of Smart Crop Guardian is its intelligent image-based skills, which let it distinguish between environmental conditions that aren't a concern and real threats from wild animals. The technology achieves a high degree of accuracy in recognizing trespassing wildlife through the use of high-resolution cameras and sophisticated image processing algorithms. Furthermore, Smart Crop Guardian interacts effortlessly with current monitoring infrastructure, resulting in a cost-effective solution that maximizes the utility of pre-existing equipment. Because of its versatility, it is a viable alternative for a wide range of agricultural contexts, from small-scale farms to large commercial enterprises. The system's user-friendly interface streamlines monitoring, allowing farmers and security professionals to handle and respond to possible threats more efficiently. In addition to providing instant protection, Smart Crop Guardian promotes sustainable farming practices by eliminating the need for conventional deterrents, which can have negative environmental consequences. The approach minimizes disturbances to the natural ecosystem while protecting important crops by accurately directing interventions. Finally, Smart Crop Guardian is a game changer in agricultural security, providing sophisticated image-based protection against wild animal invasions. Its superior technology, seamless integration, and environmentally friendly approach make it a useful asset in current farming practices, offering improved crop security and less environmental effect.

Keywords: ATMEGA328P Microcontroller, LCD, Buzzer, Power Supply Unit, Software Block Diagram.

1. Introduction

Intelligent Image-Based Protection against Wild Animal Intrusions is a game-changing breakthrough in protecting agricultural spaces from the ongoing menace of wildlife intrusions. The presence of wild animals poses a considerable difficulty in many farming situations, potentially resulting in significant crop damage and economic losses. This unique System uses cutting-edge technology to create a

comprehensive defence mechanism, integrating modern imaging sensors with complex algorithms.

The system excels at distinguishing between benign environmental conditions and actual wildlife dangers by utilizing high-resolution cameras and cutting-edge image processing. This precision is critical for allocating resources efficiently, focusing on true security problems rather than false alarms. The ability

to detect invading wildlife species precisely in real-time provides farmers and security professionals with timely information, allowing them to take immediate and targeted action. Intelligent Image-Based Protection interacts easily with current surveillance infrastructure, increasing the usability of existing systems. Because of its versatility, it is a suitable solution for a wide range of agricultural contexts, from small-scale farms to large commercial enterprises. The system's user-friendly interface accelerates the monitoring process, allowing farmers to manage and respond to possible hazards without substantial training or technical experience. Aside from its immediate protective effects, this method supports sustainable farming practices by reducing reliance on conventional deterrents that may have negative environmental consequences.

2. Detecting Animal Processes

Data Gathering and Annotation: The initial stage is to compile a dataset of animal-related photos. Annotating these photographs entails labeling each object in the image with a bounding box and identifying the class.

Model training: The YOLO model is trained using the annotated dataset. The model learns to predict bounding boxes and class probabilities based on the input photos during training.

Inference: The model is ready for inference after training. The YOLO model processes a fresh picture frame and generates a collection of bounding boxes and associated class probabilities.

Post-Processing: The YOLO model's output can include redundant or doubtful detections. After the data have been processed, the final set of detections are often refined by using Non-Maximum Suppression.

Analysis and visualization: The next stage entails visualizing the detected items on the captured photos or video frames and maybe analyzing the outcomes for a variety of applications, such as animal behavior

research, wildlife protection, or other uses.

3. Related Work

[1] it's delicate to classify animals grounded on their images and cover them more efficiently. Also, it's important to make a count of animals as they're being defunct now a days, so to save them we've to keep a note duly so that we can take important measures to save them. In being system there's only homemade checking where a mortal being need to present to keep a count of animals. [2] To address this problem, this study offers a system that uses the Internet of Things to aid in the detection of wild animal intrusions on agricultural farms by monitoring the field. Ultrasonic sensors are adopted at the field's corners, first detecting the intrusion, then capturing the image of the invader with a camera mounted on the E-vehicle embedded with a Node MCU Microcontroller that is monitoring the field. An alert message is sent to the farmer via an IoT application. The performance of the proposed system has been analyzed with respect to the captured images of the intruder and notification alert. Using this proposed model, anyone can detect any type of intrusion around the field effectively [3] the evolution of sensors has reached a completely new level thanks to image processing and IoT sensor monitoring network technology. Conflict between humans and animals threatens human lives and costs enormous amounts of resources in forested and agricultural areas. Individuals endure therefore, losing their property, animals, crops, and sporadically even their lives. Accordingly, this region should be continually watched to keep untamed life out. We have made an effort to create a system that continuously monitors the field in order to address this issue. [4] In this paper, we focus on wildlife monitoring and analysis through animal detection from natural scenes acquired by camera-trap networks. The image sequences obtained from camera-trap consist of highly cluttered images that hinder the detection of animal resulting in low-detection rates and high false discovery rates. To handle this problem, we have used a camera-trap database that has candidate animal proposals using multilevel graph cut in the

spatiotemporal domain. These proposals are used to create a verification phase that identifies whether a given patch is animal or background. [5] The conflict between wild animals and farmers is one such big issue which leads to death in both sides. In this paper, an animal detection system is used to monitor the wild animals which are intruding a village. Here thermal camera is used. It has the advantage of capturing image of wild animals even in dark. Temperature of the animal is given input of wild animal detected by the thermos graphic camera. We solve the proposed model with globally optimal guarantee. Extensive experiments on challenging well-known data sets demonstrate that our method significantly outperforms the state-of-the-art approaches and works effectively on a wide range of complex scenarios [6] This paper will discuss the present level of research and development on human-wildlife conflicts, there are bad interactions between humans and wild animals that are bad for both the resources of the human population as well as the ecosystems of wildlife. Animal detection has an impact on both human food security and animal welfare because it arises from the conflict between humans and wildlife over natural resources. [7] The idea is simple to run AI on Raspberry Pi locally to detect a wild animal and then it verifies the images. Then it sends a message through GSM module with no need of internet connection and gives an ultrasonic buzzer sound to divert a wild animal. It is trying to propose an end-to-end solution which could potentially reduce the loss of humans, animals and capitals using animal detecting system using deep convolutional Neural Networks. [8] We aim to develop an animal detection and identification system from images taken through monitoring videos captured by motion-triggered cameras, called camera traps. For these types of videos, existing approaches often suffer from low detection rates due to low contrast between the foreground animals and the cluttered background, as well as high false-positive rates due to the dynamic background. [9] When an animal is detected, an alert will be sent to another device through LoRa communication because

standard communications like Wi-Fi and GSM technologies may not be available in the remote sensing area. Computer vision using Python is utilized for image processing. Raspberry Pi development board is used where the model is deployed and live streaming is processed to detect a wild animal which will differentiate between Humans and Domestic animals. Once, a wild animal is detected an alert is sent through LoRa communication to an end-user who is monitoring the wildlife [10] To overcome this difficulty the paper suggests a classification technique using DCNN deep learning technique and ResNet152V2 algorithm, which classifies the type of animal based on their images. After classification a notification is sent using a Wi-Fi enabled IOT device through a web application software called Blynk, which controls Arduino board with smart phone via internet. In addition to this, a warning message is displayed in the LCD display, for those who don't have the application installed, which can be fixed at the Forest roads.

4. Existing System

The developed device intends to defend populated areas and agricultural areas from intrusions and attacks by wild animals. An autonomous system that can detect the intrusion of wild animals can drive them away without injury. The stated operations are carried out with the aid of a deep learning technique. An alarm sound is made after a wild animal is discovered to scare it away. To inform the appropriate authorities of the detection, an alert message is also transmitted. Thus, reducing the harmful effects of the conflict and advancing the usage of smart agriculture applications. The system's primary goal is to provide an automated method for detecting and discouraging wild animals in areas where people live and in agricultural fields. The suggested methodology uses IoT and artificial intelligence technologies to protect crops from animal encroachment. [11]

5. Proposed System

The entire system, dubbed "Intelligent Image-Based Protection from Wild Animal Intrusions," is being developed as a response to the problems brought on

by wildlife intrusions in agricultural settings. This cutting-edge system combines Python-based image processing with an Arduino Uno microcontroller, an LCD display, a buzzer, and Internet of Things integration to build a powerful defence system. [12] The Python-based image analysis module, which processes camera-captured images, is the system's central component. To discriminate between beneficial environmental conditions and serious animal threats, this module uses sophisticated algorithms. It offers timely information to prompt suitable responses by precisely identifying encroaching wildlife species in real-time. The Arduino Uno microcontroller serves as the system's central processing unit, coordinating the numerous components. When a potential threat is discovered, it gets input from the Python image analysis module and directs the output to the LCD display and buzzer, delivering immediate visual and audible alarms. This real-time feedback allows farmers and security staff to respond quickly to prevent any damage. In the device has IoT capabilities, allowing it to send email warnings when a wildlife intrusion is detected. This feature improves the system's responsiveness and enables remote monitoring, ensuring that stakeholders are kept up to date even when they are not physically present on the farm. As a user interface, the [13] LCD display offers crucial details about detected incursions, such as the species of animal and position. The system may be readily handled and monitored by farmers and security staff without the need for substantial technological skills because to the user-friendly interface. In the suggested system combines hardware elements, IoT capabilities, and intelligent image-based technologies to provide a complete defence system against wild animal intrusions in agricultural areas. This device offers a proactive method of crop protection by fusing cutting-edge image analysis with immediate notifications. The added advantage of IoT connectivity allows for rapid communication of stakeholders, which increases the system's ability to safeguard agricultural investments. Testing, Training, and Hardware Block Diagrams are shown

in Figures 1, 2, and 3.

6. Block Diagram

6.1 Software Block Diagram

6.1.1 Testing

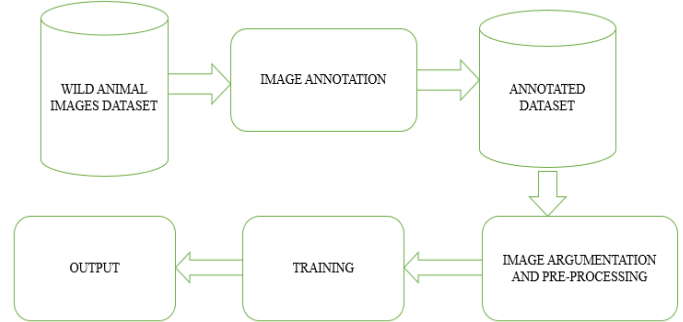


Figure 1 Testing

6.1.2 Training

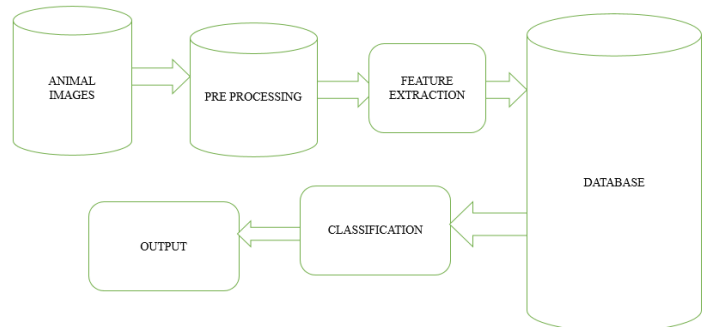


Figure 2 Training

6.2 Hardware Block Diagram

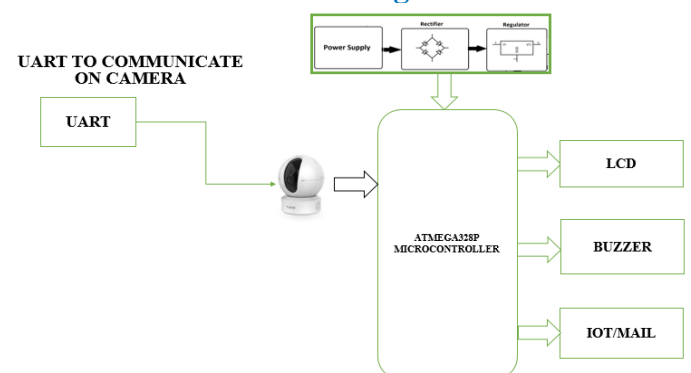


Figure 3 Hardware Block Diagram

7. Explanation

7.1 Software Explanation

The technique starts by gathering photographs with a camera strategically placed in the agricultural region.

The camera captures photos at regular intervals and provides a live feed of its surroundings. The YOLO algorithm is a well-known deep learning-based object detection system. Python is used to implement YOLO in this context. YOLO divides the image into grid cells and predicts bounding boxes and class probabilities at the same time. This means it can detect numerous items in a single neural network pass. Each frame captured by the camera is put into the YOLO algorithm for analysis. YOLO recognizes objects in the image and returns their bounding box coordinates as well as associated class probabilities. It may, for example, recognize deer, rabbits, or birds. The YOLO algorithm marks bounding boxes around the items once they have been identified to show where they are in the image. [14] These bounding boxes serve as visual markers for the animals that have been recognized. Yolo classifies items into predetermined categories in addition to localizing them. A confidence score that expresses how confident the algorithm is in its prediction is provided with each classification. To exclude detections with a low confidence level, the system can be configured using a confidence threshold. This aids in lowering false positives and concentrating on more trustworthy detections. When the YOLO algorithm recognizes an animal and the confidence threshold is exceeded, an alert is set off. This notification may be given visually on an LCD, audibly through a buzzer, or even by IoT in the mail. The system can start a response to the detected intrusion simultaneously. This can entail turning on alarms, sprinklers, or other non-lethal deterrents to prevent the animals from getting any closer. If the system has Internet of Things (IoT) capabilities, it can notify stakeholders through email when an incursion is discovered. Even when people are not there, this function assures prompt responses and improves remote monitoring.

7.2 Hardware Explanation

All components must receive steady and dependable electricity from the power supply unit. Power sources like batteries, solar panels, or a mix of the two may be used. The efficacy of the system depends on this unit's ability to operate continuously. The

ATmega328P microcontroller receives the animal detection data from Python. Usually, a serial connection or a UART port communication device is used to facilitate this communication. The ATmega328P microcontroller receives data from Python about the creatures that have been found. This information causes the alert system to go off. To get people's attention, the buzzer creates an auditory signal. When it is activated, an animal intrusion is present. A visual alert is given by the LCD, which also displays pertinent details like the kind of animal that was discovered. The system can launch a response to dissuade the animals concurrently with the alarms. To prevent further intrusion, this may entail turning on deterrents like sprinklers, alarms, or other harmless measures. The system can send email notifications to specified recipients if it has IoT capabilities. Even when they are not there, this functionality makes sure that stakeholders are instantly notified. [15]

8. Hardware Requirements

8.1 Camera



Figure 4 Camera

The detection of wild animals using a camera is shown in Figure 4 and the YOLO (You Only Look Once) algorithm to inform individuals entails a set of processes for real-time surveillance and response. It starts with the camera taking live video or photos of the monitored area. These frames are then sent into a system that employs the YOLO algorithm. YOLO, a cutting-edge object identification technique, splits the input frames into a grid and predicts bounding boxes and class probabilities for objects within each grid cell at the same time. This allows YOLO to recognize

and locate many items in a single sweep. The system then predicts bounding boxes around the discovered animals, as well as their confidence scores and class labels. This alert is delivered to the selected people through a variety of channels, including visual notifications on screens, audio alerts delivered through speakers or buzzers, and possibly even email or other IoT-enabled communication channels.

8.2 Power Supply Unit

A device or component known as a power supply unit are shown in Figure 5 delivers electrical energy to electronic systems or gadgets. It transfers electrical energy from an input source, such as a battery, wall outlet, or other power source, into a form that is appropriate for the particular requirements of the item it is powering. To maintain steady and consistent power delivery, this may entail adjusting factors such as voltage, current, and others. From small handheld devices to massive industrial machinery, power supply units are crucial parts of almost all electronic devices because they guarantee that the equipment receives the adequate and dependable electrical power it needs to operate.

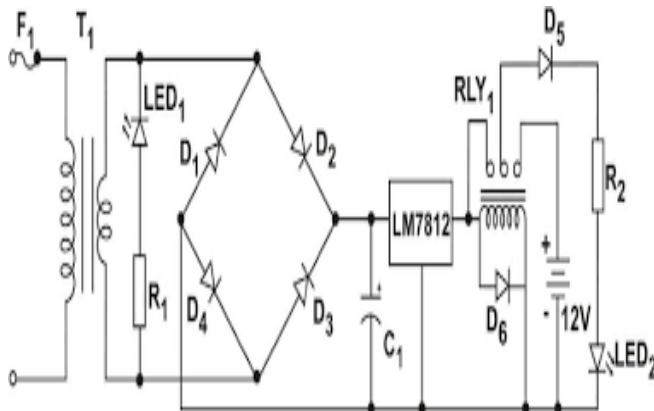


Figure 5 Power Supply Unit

8.3 8.3 ATMEGA328P Microcontroller

The discovered animal data is subsequently sent to the ATmega328P microprocessor are shown in Figure 6. A serial connection is often used for this type of communication. The ATmega328P activates the alarm system when it receives information about observed animals. When the buzzer is activated, an

audio alert is produced, signaling the presence of a detected wild animal. The LCD serves as a visual alert, providing pertinent information such as the species of animal identified. If the ATmega328P microcontroller has IoT capabilities, it can use the available internet connection to send email notifications to preset recipients. This guarantees that stakeholders are kept up to date, even if they are not present. In this integrated system combines image-based animal detection, microcontroller-based alerting, and IoT capabilities for a comprehensive defence against wild animal intrusions. It offers accurate and real-time detection by leveraging advanced detection algorithms and microcontroller capabilities. The ATmega328P microcontroller coordinates the alert system, using a buzzer and LCD display to notify stakeholders.

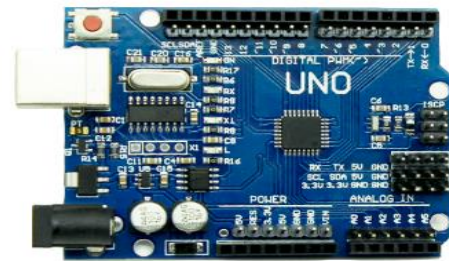


Figure 6 ATMEGA328P Microcontroller

8.4 LCD



Figure 7 LCD

A camera-based LCD (Liquid Crystal Display) are shown in Figure 7 wild animal detection system is an innovative approach for monitoring and alerting against wild animal invasions. Here's a quick rundown of how it works: An animal detection algorithm is used to process the collected frames.

This programme is intended to recognize and find animals within photos or frames. Once processed, the system returns information on detected animals, such as their location inside the frame and any additional information such as species or size. The detecting algorithm's output is subsequently relayed to the LCD panel. The LCD screen displays this information in a visual format, allowing for immediate recognition and assessment. Real-time Monitoring: The LCD screen serves as a real-time monitoring interface, providing continuous updates on any detected animal intrusions, allowing for quick response to potential threats. Simultaneously with the alert, the system can initiate a response to deter the animals, which may include activating deterrents such as sprinklers, alarms, or other non-harmful measures.

8.5 Buzzer



Figure 8 Buzzer

When a wild animal is discovered, a sensor sends a signal to the alarm system to start the warning procedure. When the signal is received, the system sounds a buzzer as shown in Figure 8. Anyone around is immediately alerted by the buzzer's loud, attention-grabbing sound. The sound that is released acts as a loud, distinct warning of imminent wildlife entry. In order to ensure that locals are instantly aware of the situation, this audio alert is essential. The technology is intended to continuously scan the area for any indications of animal movement, making sure that any fresh intruders are quickly found and reported.

8.6 Node MCU

Sensors such as motion detectors, infrared sensors, or image-based algorithms are used in the system to monitor the environment for signals of wild animal activity. Node MCU is a microcontroller board with

an ESP8266 Wi-Fi module that allows it to connect to the internet. It functions as the alert system's primary processing unit. When the sensors identify a wild animal, they transmit a signal to the Node MCU board, alerting it to a possible intrusion. Because NodeMCU is Wi-Fi-enabled, it can connect to the internet and deliver notifications to external servers or services. When a wildlife intrusion is detected, the NodeMCU board is programmed to send an email notice. This entails setting up the board to communicate with an email server via secure protocols such as SMTP (Simple Mail Transfer Protocol). Critical details like the species of animal discovered, the moment it was discovered, and possibly even a picture or video feed from the monitoring system are frequently included in the email message. Other reaction mechanisms, such as setting off alarms, turning on lights, or sending notifications to mobile devices through messaging platforms, may also be triggered concurrently with the email message by the system. The system is made to continuously scan the surrounding area for any indications of animal movement, making sure that any fresh invasions are quickly found and reported. The system's settings may be programmable, enabling users to alter the sensors' sensitivity, the details of email warnings, and other aspects to suit their own requirements.

Conclusion

Finally, Smart Crop Guardian is at the cutting edge of agricultural security, revolutionizing how we defend our crops from wild animal invasions. This system provides a complete and proactive defence mechanism by leveraging the capabilities of intelligent image-based technology. The seamless integration of powerful sensors and sophisticated algorithms guarantees accurate and real-time detection of benign environmental elements and genuine hazards. Smart Crop Guardian's versatility, which allows it to effortlessly integrate with current surveillance infrastructure, makes it a viable solution for a wide range of agricultural settings. Its easy-to-use interface and high precision limit the risk of false alarms, allowing for effective resource allocation and

targeted intervention when necessary. And it is significant advancement in agricultural security. Its mix of cutting-edge technology, agility, and environmental sensitivity makes it a potent instrument for protecting agricultural investments and maintaining long-term farming practices. This method exemplifies the power of technology to address complex agricultural difficulties, ultimately improving food security and promoting harmonious cohabitation between farming communities and the natural world.

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