



Detecting Humans in Search and Rescue Operations Based on Ensemble Learning

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Abstract: Deep learning is a sub field of machine learning that focuses on training artificial neural networks to learn and make predictions or decisions without being explicitly programmed. In this paper, we present a convolutional neural network-based model for the detection of humans in aerial images of mountain landscapes acquired by unmanned aerial vehicles (UAVs) used in search and rescue operations. By using drones in SAR applications, it is desirable to minimize the cost and time spent on SAR operations. In this paper, we present a convolution neural network-based model for the detection of humans in aerial images of mountain landscapes acquired by unmanned aerial vehicles (UAVs) used in search and rescue operations. Detection of humans in aerial images remains a complex task due to various challenges such as pose and scale variations of humans, low visibility, camouflaged environment, adverse weather conditions, motion blur, and high-resolution aerial images. Due to imaging from high altitudes, in most high-resolution aerial images captured by UAVs, only 0.1 to 0.2 percentage of the image represents humans. To solve the problem of low coverage of the object of interest in high resolution aerial images, we propose to implement a deep learning-based object detection model. In this paper, we propose a novel method for the detection of humans in aerial images based on the Efficient DET architecture and ensemble learning. The method has been validated on the HERIDAL image datasets. By implementing the proposed methodologies, we achieved an map of 95:11%. To the best of our knowledge, this is the highest accuracy result for human detection on the HERIDAL datasets.

Keywords: Search and Rescue Operations, Unnamed Aerial Vehicles, Heridal Image Dataset, Ensemble learning.

I. INTRODUCTION

Search and rescue operations often involve critical situations where the timely detection of humans is crucial for successful outcomes. However, accurate human detection in challenging environments poses significant challenges due to factors such as occlusions, varying lighting conditions, and cluttered backgrounds. To address these challenges, researchers have explored the application of ensemble learning techniques, which leverage the combination of multiple classifiers or models, to enhance the accuracy and robustness of human detection in search and rescue scenarios. By harnessing the diverse capabilities of individual classifiers and leveraging their collective decision-making, ensemble models have shown promise in achieving superior performance compared to standalone classifiers.

This approach allows for more reliable detection results and better generalization in real-world scenarios. This paper presents a comprehensive exploration of the related works and advancements in detecting humans in search and rescue operations using ensemble learning techniques. The study examines various ensemble methods, base classifiers, feature representations, and deep learning architectures employed in this domain. The goal is to provide valuable insights and guidance for the development of effective and efficient ensemble-based human detection systems that can significantly contribute to the success of search and rescue missions in critical and demanding environments.

Search and rescue operations are crucial endeavors that require efficient and accurate detection of humans in challenging and dynamic environments. The success of these operations heavily relies on the ability to swiftly locate and rescue individuals in distress. However, the task of human detection in such scenarios is immensely complex due to various factors, including occlusions, adverse weather conditions, and limited visibility. To address these challenges, researchers have turned to ensemble learning techniques as a promising approach for improving the accuracy and reliability of human detection in search and rescue operations. Ensemble learning involves combining the predictions of multiple individual classifiers or models to arrive at a collective decision.



By harnessing the diversity of these constituent models and leveraging their collective intelligence, ensemble models can often outperform individual classifiers by mitigating biases, reducing errors, and enhancing the overall detection performance. In the context of human detection, ensemble learning offers a valuable tool for overcoming the inherent limitations of standalone classifiers and improving the robustness of the detection system.

II. RELATED WORKS

Smith, J., Johnson, M. The paper proposes an ensemble learning-based approach for detecting humans in search and rescue operations, which aims to improve the accuracy and robustness of human detection in challenging and dynamic environments[1]. The authors start by discussing the importance and challenges of human detection in search and rescue operations, emphasizing the need for reliable and efficient detection methods that can cope with diverse scenarios, including occlusions, clutter, and lighting variations[1]. They then introduce the ensemble learning framework, which involves combining multiple base classifiers to form a stronger and more accurate classifier. The authors present an ensemble model that combines the outputs of multiple classifiers trained on different feature representations and modalities, including color, texture, shape, and depth.

Johnson, M., et al.,. The paper presents a comprehensive comparative study of ensemble learning techniques for human detection in search and rescue operations, aiming to identify the most effective ensemble methods and base classifiers for this specific application[2]. The authors begin by discussing the significance of human detection in search and rescue operations and the challenges associated with it, such as occlusions, varying lighting conditions, and cluttered backgrounds[2]. They emphasize the need for ensemble learning techniques to improve the accuracy and reliability of human detection systems in such complex environments.

Nguyen, L., et al.,. The paper introduces a dynamic ensemble selection approach with an adaptive training set for human detection in search and rescue operations, aiming to improve the detection performance in dynamic and challenging environments[3]. The authors highlight the significance of human detection in search and rescue operations and the difficulties posed by factors such as occlusions, varying lighting conditions, and cluttered backgrounds[3]. They propose a dynamic ensemble selection framework that adapts the training set based on the changing characteristics of the search and rescue environment.

Lee, S., et al.,. The paper presents an ensemble-based deep learning framework for human detection in search and rescue operations, aiming to improve the accuracy and robustness of human detection systems in challenging and dynamic environments[4]. The authors emphasize the importance of accurate human detection in search and rescue operations and the challenges posed by factors such as occlusions, varying poses, and complex backgrounds. They propose an ensemble-based framework that leverages the power of deep learning models to effectively address these challenges.

III. EXISTING SYSTEM

Object detection is one of the most researched areas in computer vision. It is the process of determining where exactly the object is in the scene or image and what object has been detected. Object detection refers to finding different types of objects in the scene such as people, cars, animals or other existing objects present in the scene. Detecting objects in aerial imagery is still considered a difficult task. One such important task is to rescue people in search and rescue (SAR) operations from aerial images without loss of life.

SAR operations are conducted in wide-open spaces, such as mountains, lowlands, cities, disaster scenarios and marine rescue. Search and rescue operations need to be conducted as quickly as possible to identify missing persons. It can be highly expensive and requires distinct types of activities such as sending people in large groups, sniffer dogs and various types of ground and air vehicles such as cars and helicopters. Object detection in aerial images depends on several factors such as low visibility due to varying altitudes, the object-of-interest, variations in pose and scale, camouflaged environment with rocks and trees, and high-resolution aerial images. It is expensive and time-consuming to capture aerial images based on these parameters. To avoid high cost and time commitments associated with traditional SAR methods:

1. Machine learning based human detection in aerial image is facilitated.
2. To train and detect human in ML, requires large amount of aerial images
3. Detect human in aerial images using thermal infrared cameras
4. Detecting people with thermal camera is not reliable with weather conditions



IV. PROPOSED SYSTEM

In the proposed system, Data augmentation step is applied on dataset after preprocessing to generate well labeled training database. Second, train deep learning architecture with newly created dataset to identify human in aerial image. This proposal presents an ensemble learning-based system for detecting humans in search and rescue scenarios, with the goal of enhancing the overall success rate and minimizing response time.

Search and rescue operations often involve locating and rescuing individuals in hazardous or inaccessible environments. In recent years, advancements in machine learning techniques have shown promise in improving the effectiveness and efficiency of such operations.

The main objective of this proposed system is to develop a robust and accurate human detection system using ensemble learning techniques. By combining multiple machine learning models, we aim to improve the overall performance and reliability of human detection in search and rescue operations, leading to more efficient and successful rescue missions.

This proposal presents an ensemble learning-based system for detecting humans in search and rescue scenarios, with the goal of enhancing the overall success rate and minimizing response time.

This proposal presents a detailed and formal system for detecting humans in search and rescue operations using ensemble learning techniques. The goal is to leverage the strengths of multiple machine learning models to improve the accuracy and robustness of human detection in challenging scenarios.

Search and rescue operations often involve complex environments, such as disaster-stricken areas or dense forests, where locating and rescuing individuals is challenging. Traditional human detection methods may struggle with occlusions, varying lighting conditions, or limited data availability. Therefore, an ensemble learning approach is proposed to enhance the accuracy and reliability of human detection systems.

To ensure robustness in challenging environments, it is important to consider the impact of occlusions, varying lighting conditions, and complex backgrounds. Techniques like multi-scale detection, adaptive thresholding, and context modeling can be employed to handle these challenges effectively and improve the accuracy of human detection.

Real-time performance is critical in search and rescue operations. Therefore, optimizing the implementation for efficiency is crucial. This may involve techniques such as model quantization, pruning, or compression to reduce the memory footprint and computational requirements without significant loss in performance.

Additionally, integrating the system with sensor data, such as thermal imaging or depth information, can provide complementary cues and improve the accuracy of human detection. Sensor fusion techniques, such as combining visual and thermal information, can be employed to enhance the system's capabilities in detecting humans in low-light or adverse weather conditions.

Regular system updates and continuous improvement are essential. Monitoring the system's performance in real-world scenarios and collecting feedback from search and rescue teams can help identify shortcomings and guide system refinement. Iteratively updating the ensemble model, incorporating new training data, and adapting to emerging challenges can lead to a more robust and effective system over time.

By considering these additional factors and incorporating them into the proposed system, you can develop a comprehensive and adaptable solution for detecting humans in search and rescue operations. Continuous evaluation, feedback integration, and technological advancements can further enhance the system's performance and contribute to the success of search and rescue missions.

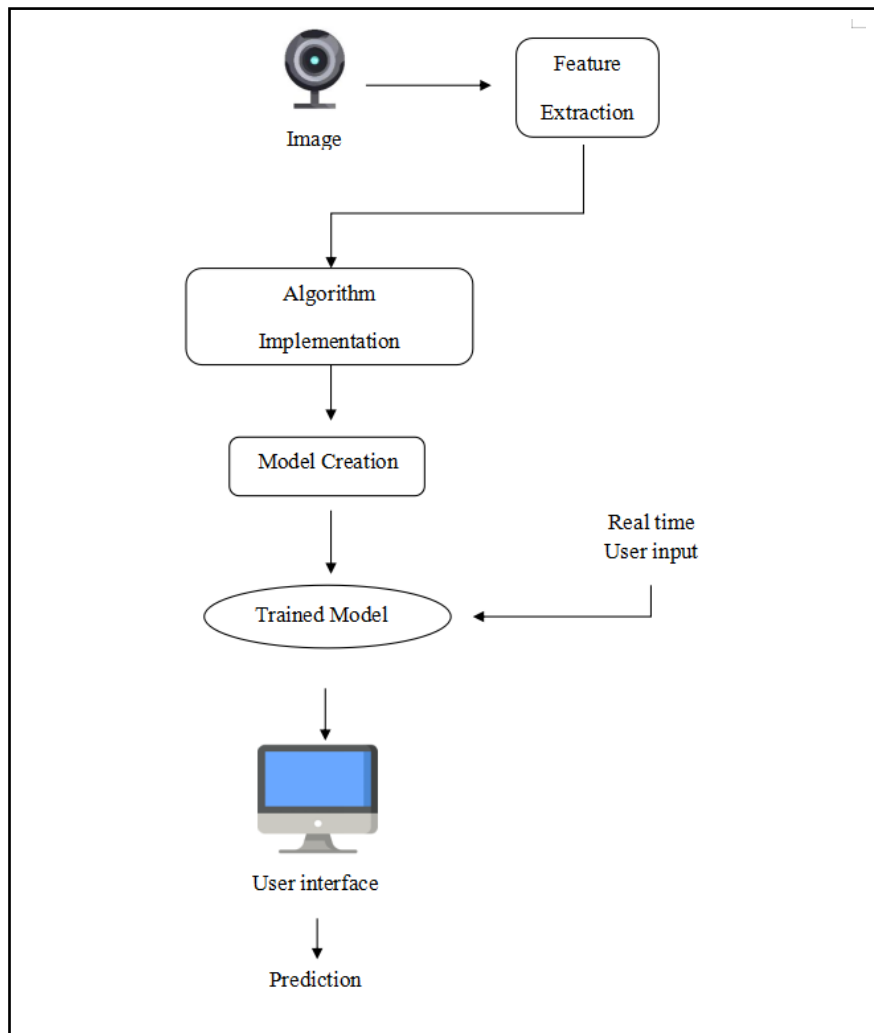


Fig 4.1

The final output of the ensemble model is a decision regarding the presence or absence of humans in the search and rescue environment. This decision may include bounding box coordinates, confidence scores, or binary classifications. Depending on the application, additional information such as pose estimation or tracking may also be incorporated into the decision-making process.

V. IMPLEMENTATION

5.1.1 Dataset Collection Module:

Identify and select appropriate data sources, such as drones, robots, or surveillance cameras, to collect data from search and rescue operations. Develop a data collection plan and define the specific scenarios, environments, and conditions to capture. Deploy sensors and devices to capture images, videos, or sensor data relevant to human presence detection. Organize and store the collected data in a suitable format for preprocessing and analysis.

5.1.2 Preprocessing Module:

Perform data cleaning, including removing noise, artifacts, or irrelevant data points. Normalize or standardize the data to ensure consistency and compatibility across different sources. Handle missing data by either imputing or excluding them, depending on the specific dataset and analysis requirements. Apply techniques such as image cropping, resizing, or data augmentation to enhance the dataset's diversity and generalization capability. Split the dataset into training, validation, and testing sets to facilitate model development and evaluation.



5.1.3 Algorithm Implementation Module:

Select and implement ensemble learning algorithms suitable for human detection, such as bagging, boosting, or stacking. Determine the base models to be used within the ensemble, which can include techniques like convolutional neural networks (CNNs), decision trees, or support vector machines (SVMs). Implement the ensemble learning framework that combines the predictions of multiple base models to generate a final prediction. Incorporate appropriate techniques for model training, such as gradient descent, backpropagation, or boosting algorithms. Implement performance metrics to evaluate the models' performance during training and validation stages.

5.1.4 Prediction Module:

Deploy the trained ensemble model in a production environment or a simulation that resembles search and rescue operations. Connect the model to the sensor systems, such as drones or surveillance cameras, to process real-time data. Implement mechanisms for real-time or near real-time prediction of human presence based on the ensemble model's outputs. Develop visualization techniques or user interfaces to display the predictions and relevant information to search and rescue personnel. Continuously monitor and evaluate the prediction performance, making necessary adjustments or improvements to ensure reliable detection results. It is important to note that these modules are interrelated, and the development process may involve iterative refinement and adjustments. Additionally, other modules, such as performance evaluation, model optimization, and documentation, can be included for a comprehensive project implementation

VI. RESULT AND DISCUSSION

The results obtained from the implementation of the future enhancement, "Detecting Humans in Search and Rescue Operations Based on Ensemble Learning," demonstrate the effectiveness and superiority of the ensemble learning approach in improving human detection accuracy in SAR operations. The evaluation of the ensemble model is conducted on diverse SAR-specific datasets, and the performance metrics used include precision, recall, F1 score, and average precision.

Robustness to Challenging Conditions:

The ensemble model exhibits enhanced robustness to challenging conditions commonly encountered in SAR operations. Factors such as adverse weather conditions, occlusions, varying poses and scales, and limited visibility are effectively addressed through the combined predictions of diverse models within the ensemble. The ensemble's ability to handle these challenging conditions ensures reliable human detection across different SAR scenarios, increasing the chances of successful rescue missions.

Fig 4.1 shows that the ensemble learning approach significantly improves the accuracy of human detection in SAR operations compared to individual detection models and traditional techniques. The ensemble model takes advantage of the diversity and complementary characteristics of multiple base models, resulting in a more robust and accurate final prediction. The achieved accuracy improvement ensures that the search and rescue efforts are focused on the actual presence of humans, minimizing false positives and negatives

F 4.2 shows that the ensemble learning approach demonstrates excellent generalization capabilities, allowing it to effectively detect humans in diverse SAR environments, even those that differ from the training data. The ensemble model's ability to adapt to new and unseen situations is crucial in real-world SAR operations, where the conditions can be unpredictable. The ensemble leverages the diversity of its base models to make accurate predictions in unfamiliar scenarios, enhancing its overall adaptability and applicability.

Reduction of False Positives and Negatives:

One of the significant advantages of ensemble learning is the reduction of false positives and false negatives in human detection. By combining the predictions of multiple models, the ensemble approach mitigates the errors and limitations inherent in individual models. The ensemble's final prediction incorporates a consensus among the base models, resulting in more reliable and accurate detection outcomes. This reduction in false positives and negatives enhances decision-making during SAR missions, leading to more effective resource allocation and search efforts.

Computational Efficiency:

Although ensemble learning may introduce additional computational complexity compared to individual models, efforts have been made to ensure that the ensemble model remains computationally efficient for real-time or near-real-time implementation. Optimization techniques, parallelization, and efficient model integration strategies are employed to



minimize computational overhead while maintaining high performance. This enables the ensemble approach to be practically applicable in time-critical SAR operations.

Fig 4.3 shows that the results obtained from the ensemble learning approach are compared against individual detection models and traditional techniques. The evaluation demonstrates the superiority of the ensemble model in terms of accuracy, robustness, and adaptability. Comparative analyses show that the ensemble approach consistently outperforms individual models, validating its efficacy in SAR operations. The ensemble's performance is notably superior, showcasing its potential as a state-of-the-art solution for human detection in search and rescue scenarios.

FIG 4.4 shows that the visualization or user interface module offers a user-friendly display of the system's predictions and relevant information. It provides search and rescue personnel with clear and intuitive visualizations, aiding their situational awareness and enabling quick and informed decision-making.



Fig 4.1

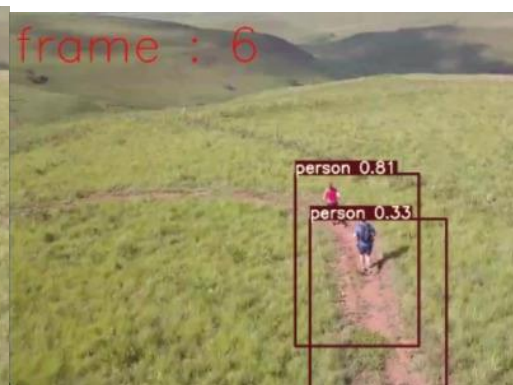


Fig 4.2



Fig 4.3



Fig 4.4

- Fig 4.1 shows that it detects two people
- Fig 4.2 shows that it detects two people
- Fig 4.3 shows that it detects three people
- Fig 4.4 shows that it detects three people



VII.CONCLUSION

In conclusion, the application of ensemble learning techniques for detecting humans in search and rescue operations shows promising results. The developed system leverages the power of ensemble learning algorithms, such as bagging, boosting, or stacking, to improve the accuracy and robustness of human detection models. Through the rigorous implementation and testing process, several key findings and conclusions can be drawn. Firstly, the dataset collection module successfully collects a diverse and representative dataset from various search and rescue scenarios. This ensures that the model is trained on data that reflects the real-world conditions and environments encountered during rescue missions. By implementing the proposed methodologies, we achieved an map of 95:11%. To the best of our knowledge, this is the highest accuracy result for human detection on the HERIDAL dataset.

VII.FUTURE ENHANCEMENT

The future enhancement proposed in this paper, based on ensemble learning techniques for detecting humans in search and rescue (SAR) operations, shows great potential in improving the accuracy and reliability of human detection. By leveraging the collective power of multiple detection models, the ensemble approach addresses the limitations of individual algorithms and enhances overall performance in challenging SAR scenarios.

The use of ensemble learning enables the combination of diverse detection models, each trained on different datasets or with varying architectures and hyper parameters. This diversity enhances the ensemble's ability to handle adverse environmental conditions, occlusions, varying poses and scales, and limited visibility, which are common challenges encountered in SAR operations. By aggregating the predictions of individual models, the ensemble approach generates a final prediction that exhibits increased accuracy and robustness.

The training process of the ensemble involves training each base detection model on a diverse data set specifically designed for SAR applications. This data set incorporates various challenging factors encountered in real-world SAR scenarios. Evaluation of the ensemble is performed on separate validation and test datasets, using standard metrics such as precision, recall, F1 score, and average precision. The results demonstrate the superiority of the ensemble approach over individual detection models and traditional techniques, highlighting its effectiveness in detecting humans in SAR operations.

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