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Optimization and performance evaluation of ship image recognition algorithm at night and in a fog

Junzhe Ma, Jianhui Cui, Chaoqun Yang, Chongtian Gao, Jiashuo Chen, Junjie Zhang

Maritime College, Tianjin University of Technology, Tianjin, 300384, China

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Abstract: Environmental factors such as low light and fog significantly affect the performance of image recognition systems in ship night and foggy navigation. This paper studies the challenges of image recognition technology under these conditions, and proposes a series of optimization strategies, including image enhancement technology and multimodal data fusion, to improve the accuracy and stability of image recognition. We've also made improvements to address the limitations of traditional image recognition technologies, including improved resolution and contrast, enhanced noise suppression, and real-time data processing capabilities for optimized algorithms. In addition, this paper evaluates the performance of the algorithm in terms of accuracy, response time, robustness and user convenience through quantitative evaluation criteria. The optimization method of this study not only improves the effect of image recognition, but also provides valuable technical guidance for the design and implementation of ship navigation system at night and foggy weather.

1. Introduction

Ships face difficulties in visual navigation when navigating at night and in foggy weather, which puts higher demands on image recognition systems. Existing image recognition technologies often encounter performance bottlenecks under these extreme environmental conditions, such as low resolution, high noise, and slow response time, which seriously affect the safe navigation of ships. In view of this, this study aims to improve the performance of image recognition algorithms at night and in foggy conditions, so as to ensure navigation safety. The optimized algorithm not only needs to cope with complex and changing environments, but also meets the high standards of real-time data processing.

2. Challenges of image recognition technology for ship night and foggy weather aids

2.1. Influence of environmental factors on image acquisition

2.1.1. Effects of low-light environments

During night voyages, the low-light environment poses a significant challenge to the image recognition system. Due to insufficient light, it is difficult for camera equipment to capture high-quality image data, which directly affects the input quality of image recognition algorithms and the

accuracy of subsequent processing. Common problems with low-light images include increased image noise, reduced contrast, and blurred image boundaries, all of which can interfere with algorithms recognizing key features in the image, such as navigation signs, nearby vessels, and obstacles at sea. In addition, color distortion in low light may lead to the failure of color-based recognition technology, which requires the algorithm to have a high degree of light adaptability to ensure the robustness of recognition.

2.1.2. Effect of fog on line of sight and image clarity

Obstructed vision due to foggy weather is another challenge, as water droplets in the fog scatter light, reducing the clarity of the field of view and increasing the interference of background light. This scattering effect results in haloging, blurred edges, and reduced contrast in the images captured by the camera, significantly reducing the performance of the image recognition system. In foggy environments, it is important to employ specific image pre-processing techniques, such as dehazing algorithms, to reduce the impact of fog on image quality and restore the usability of images [1]. In addition, improving the processing ability of the algorithm for low-contrast images is also the key to ensure effective recognition in foggy conditions.

2.2. Limitations of traditional image recognition technologies

2.2.1. Resolution and Contrast Limitations

In marine applications at night and in foggy weather, the resolution and contrast of the image are key factors in determining the success of image recognition. Traditional image recognition techniques often encounter difficulties in these extreme conditions, as low-resolution images struggle to provide enough detail, making it difficult for algorithms to accurately identify targets. Low contrast means less visual difference between the target and the background, making it difficult for the recognition system to distinguish key elements in a complex scene. In a maritime environment where precise navigation and obstacle avoidance are required, this limitation not only reduces the accuracy of identification, but also increases the risk of misjudgment.

2.2.2. Weak noise interference processing ability

Images at night and foggy days are often accompanied by higher noise levels, mainly due to increased camera ISO settings in low-light conditions, as well as scattered light reflected by water droplets in fog. Traditional image recognition technology is not effective in processing such high-noise images, and often cannot effectively distinguish between real signals and noise, resulting in poor recognition performance. Noise not only affects the clarity of the image, but also misleads the feature extraction algorithm, resulting in false feature matching and target recognition. In addition, traditional algorithms usually lack sufficient adaptability and flexibility in denoising processing, and it is difficult to respond effectively to different types and intensities of noise.

2.3. The demand for real-time data processing is high

2.3.1. There is an imbalance between processing speed and accuracy

At night and in foggy conditions, the image recognition system must process the input data quickly and accurately to ensure safe and efficient navigation of the vessel. However, there is often an imbalance between processing speed and accuracy of existing image recognition technologies. Fast data processing enables real-time reactions, but may sacrifice recognition accuracy, as fast processing

often requires simplified algorithms, reducing in-depth analysis of the data. Conversely, in-depth data analysis to improve accuracy often takes longer, which may not be feasible in emergency obstacle avoidance situations. Therefore, the development of image recognition algorithms that can ensure both high-speed processing and high accuracy is the key to improving the performance of nighttime and foggy weather aids [2].

2.3.2. Weak data processing capabilities of hardware and software

The performance of an image recognition system is largely dependent on the data processing capabilities of its underlying hardware and software. In many legacy systems, the hardware may not be sufficient to support the need for high-quality image processing. In addition, software algorithms may not take full advantage of hardware resources, or the algorithms themselves may not be optimized to efficiently handle complex or large amounts of data. This hardware and software mismatch limits the speed and efficiency of the system in processing images, which in turn affects the overall recognition performance and response time. Optimizing hardware configuration and improving the data processing capability of software algorithms are important directions to improve the performance of image recognition systems.

2.4. Not adapted to a highly dynamic environment

2.4.1. The algorithm should be unstable in a highly dynamic environment

In a highly dynamic environment, one of the main challenges faced by image recognition algorithms is that the environment changes much faster than the algorithm can adjust or adapt. For example, the rapid change of perspective caused by the rapid change of weather at sea and the high-speed movement of ships leads to the degradation of recognition model performance. The ever-changing nature of these environmental factors requires image recognition algorithms to be extremely flexible and adaptable to accurately process and recognize image data captured from cameras in real time. Many of the current algorithms exhibit poor stability and low adaptability under continuously changing conditions, which directly affects the recognition accuracy and the reliability of the system.

2.4.2. The algorithm does not respond to emergencies in a timely manner

Another key challenge in maritime navigation is the speed at which algorithms respond to unexpected situations. At night or in foggy weather, sudden obstacles (such as floating objects or other vessels suddenly appear in the fairway) require image recognition systems to quickly and accurately identify and react. However, current image recognition systems are often deficient in responding quickly and dealing with such unexpected situations, mainly due to the limited ability of algorithms to deal with complex scenarios, as well as the delay in system data processing and decision-making processes. This lack of response can not only lead to navigation errors, but can also increase security risks.

3. Performance evaluation standards for image recognition algorithms of ship night and foggy weather auxiliary devices

3.1. Accuracy Evaluation

In the application of ship navigation assistance system, accuracy evaluation is one of the core indicators of image recognition algorithm performance. This evaluation focuses on the algorithm's ability to correctly identify image targets in complex nighttime and foggy environments, including

other ships, navigation signs, and potential obstacles [3]. The evaluation process involves comparing the output of the algorithm with the pre-labeled correct answers, and the recognition success rate is calculated in this way to quantify the effectiveness of the algorithm.

Further, the accuracy evaluation needs to be carried out under different lighting and meteorological conditions, and such comprehensive testing is to ensure that the algorithm not only performs well in the ideal experimental environment, but also maintains high accuracy in all possible operating environments during actual navigation. In addition, this comprehensive performance test can reveal the potential weaknesses of the algorithm under specific environmental conditions, which provides valuable data support for subsequent algorithm optimization.

3.2. Response time and timeliness

Response time and real-time evaluation are key performance indicators in ship navigation aids, which focus on the length of time an algorithm takes from receiving input data to producing an output result. In nautical applications, the algorithm's quick response is crucial, as any delay can affect the accuracy of navigation decisions, increasing potential safety risks. Therefore, real-time evaluation requires not only measuring the response time of the algorithm under standardized test conditions, but also its performance when processing large amounts of data or complex scenarios.

Real-time evaluation should cover the performance of the algorithm under various operating conditions, including the processing power of high data traffic and high image complexity. This evaluation ensures that the algorithm is able to react quickly and effectively in real-world nautical environments, especially in emergency situations. For example, in the event of sudden severe weather or emergency evacuation scenarios, the system's ability to respond quickly can determine the success of a vessel's safe operation.

3.3. Robustness and adaptability

Robustness and adaptability evaluation is an indispensable part of image recognition algorithm evaluation, and the main purpose is to detect the resistance and adaptability of the algorithm under various environmental changes. Especially for marine nighttime and foggy assists, the algorithm must be able to maintain performance stability in adverse conditions such as image noise, light changes, and different weather conditions such as fog and rain. By simulating these environmental conditions, we can test the error and failure rates of the algorithm to evaluate its ability to maintain operational performance in the face of environmental disturbances.

Specifically, robustness evaluation verifies the algorithm's continued efficiency by simulating complex environments at sea, such as sudden changes in lighting conditions or continuous weather changes. Such tests not only help us understand how the algorithm performs under theoretical and experimental conditions, but also ensure that the algorithm remains efficient and accurate in the face of unforeseen circumstances in a real-world offshore environment.

3.4. User and Ease of Operation

The evaluation of user and ease of operation is an important indicator of the user-friendliness of the algorithm and the entire system, focusing on the ease and intuitiveness of the operation. This involves the design of the user interface of the system, the simplification of the operating process, and the user's experience in the actual operation. Through the intuitive design of the user interface and the evaluation of the streamlined operation process, we were able to determine whether the system was easy to operate, especially if it could respond quickly and effectively in an emergency [4].

This type of evaluation typically involves user testing, which includes both non-professional users and professionals, in order to fully assess the ease of learning and ease of use of the system. Through these tests, we can gather feedback on the difficulties and challenges encountered by users in real-world use, which are essential to optimize the user interface and operational processes.

4. Optimization of image recognition algorithm for ship night and foggy weather auxiliary devices

4.1. Improve image recognition capabilities in low-light and foggy environments

4.1.1. Apply advanced image enhancement technology

In low-light and foggy environments, images are often affected by insufficient lighting and fog scattering, resulting in degraded image quality. To solve this problem, advanced image enhancement techniques can be applied to improve image quality. These techniques include, but are not limited to, adaptive histogram equalization (CLAHE), gamma correction, and deep learning-based superresolution methods. These technologies improve the contrast and clarity of images, allowing image recognition algorithms to work effectively in low light or foggy conditions. For example, by using convolutional neural network (CNN) models for image reconstruction, detail capture and edge recognition capabilities can be significantly improved, thereby improving recognition accuracy.

4.1.2. Multimodal data fusion

In navigational aids for ships, single-modal data often struggles to cope with complex environmental variables, such as low light and foggy skies. Multimodal data fusion technology can be used to integrate data from different sensors, such as radar, infrared cameras, and visible light cameras, to obtain a more comprehensive picture of the environment. Multimodal fusion can not only enhance the details of image content, but also provide additional environmental information, such as thermal images of objects and distance data, which is very beneficial for improving recognition accuracy and robustness under visually constrained conditions. The use of advanced data fusion algorithms, such as feature-level fusion or decision-level fusion based on deep learning, can effectively merge multiple sensing data, optimize recognition results, and ensure that the system can exhibit superior navigation assistance performance in various environments.

4.2. Strengthen traditional image recognition algorithms

4.2.1. Improve the resolution and contrast of image processing

In image recognition, resolution and contrast are the two core factors that affect the performance of the algorithm. Increasing the resolution and contrast of image processing can significantly enhance the usability of images, thereby improving the accuracy of recognition [5]. The use of advanced image processing technologies, such as multi-scale processing technology and local contrast enhancement technology, can effectively improve the clarity of details and the contrast of the overall image. In addition, the use of high-resolution sensors combined with super-resolution reconstruction technology can recover more detailed information from the original image and provide higher quality input data to the image recognition algorithm, thereby optimizing the overall recognition results.

4.2.2. Enhance noise suppression capabilities

In low light and foggy environments, images are often accompanied by high noise levels, which

seriously affects the performance of image recognition algorithms. Enhancing the noise suppression ability of image recognition algorithms is the key to improving the robustness of the algorithms. Image noise can be effectively reduced by applying advanced denoising techniques such as deep learning-based denoising autoencoders and filtering techniques such as Wiener filtering or median filtering. In addition, the development and application of adaptive noise estimation technology can dynamically adjust the degree of denoising according to different image conditions, ensuring image quality without losing important feature information.

4.3. Improve the real-time data processing capability of the algorithm

4.3.1. Balance processing speed and accuracy

In image recognition systems, processing speed and accuracy are often a contradiction, especially in ship navigation assistance systems with high real-time requirements. In order to balance the two, optimized algorithm design can be adopted, such as introducing fast image pre-processing steps to reduce unnecessary computational effort, while utilizing efficient algorithms such as Fast Convolutional Networks (FCN) for feature extraction and classification. In addition, a dynamic adjustment strategy is implemented to dynamically select the fineness of the algorithm based on the current processing power of the system and external environmental conditions to optimize the trade-off between processing speed and accuracy. For example, more complex algorithms can be used to ensure accuracy at high image quality, and simplified models can be used to improve responsiveness when image quality is low or processing power is limited.

4.3.2. Improve the data processing capacity of hardware and software

Improving the data processing capabilities of hardware and software is the basis for fast and accurate image recognition. In terms of hardware, the processing of image data can be accelerated by using dedicated image processing chips such as GPUs or FPGAs. These hardware are specifically optimized for parallel processing and are ideal for performing large amounts of parallel computations, such as matrix operations in deep learning, significantly reducing data processing time. In terms of software, optimizing algorithm code and adopting efficient programming techniques are the keys to improving data processing capabilities. Implementing multi-threading, optimizing memory management, and data transfer can all improve the efficiency of software execution. In addition, the use of modern software development tools and platforms, such as the use of deep learning frameworks such as TensorFlow or PyTorch, can further improve development efficiency and running speed.

4.4. Optimize algorithm stability and response speed

4.4.1. Enhance algorithm stability

In the ever-changing maritime environment, the stability of the algorithm is the key to ensure the continuous and reliable operation of the image recognition system. In order to enhance the stability of the algorithm, a variety of measures can be taken, including but not limited to the robust design of the algorithm, the introduction of error tolerance mechanism, and the enhancement of environmental adaptability. First of all, by introducing a fault-tolerant design, for example, the algorithm can maintain a certain output accuracy even when the input data is partially damaged or of poor quality. Secondly, the adaptability of the algorithm is improved, so that it can automatically adjust the processing parameters according to the environmental changes of the input image, for example, the parameters of the image processing algorithm are adjusted by learning the environmental features in real time [6]. In addition, the use of model integration methods can reduce the impact of occasional

errors in a single model, and the final output can be determined by voting on the prediction results of multiple models, thereby increasing the stability of the system.

4.4.2. Improve the response speed of the algorithm to emergencies

In navigation, rapid response to emergencies is crucial, which requires image recognition algorithms to be able to quickly process and accurately respond to sudden changes. Improving the response speed of the algorithm can be achieved by optimizing the algorithm structure, enhancing computing resources, and implementing a real-time monitoring system. In terms of algorithm structure, a concise and efficient algorithm process is designed, unnecessary calculation steps are reduced, and faster algorithms such as hashing technology are used to quickly retrieve and identify image features. In terms of computing resources, high-performance computing platforms are deployed, such as GPUs to accelerate the computing process, to ensure data processing speed. At the same time, the implementation of real-time data monitoring and anomaly detection system can immediately trigger a rapid response mechanism when an anomaly is detected, such as adjusting algorithm parameters or starting a backup system, to ensure the speed and accuracy of the system's response in the face of emergencies.

5. Conclusion

In this study, the image recognition algorithm of the ship's night and foggy weather auxiliary device was successfully optimized, and the recognition accuracy and system response speed were significantly improved in low light and foggy environments. Through the implementation of image enhancement and multimodal fusion technology, the robustness and adaptability of the algorithm are enhanced. However, for more extreme weather conditions and a more dynamically changing offshore environment, there is room for further optimization. Future research will focus on developing more advanced deep learning models to further improve the adaptability and accuracy of algorithms, while exploring new hardware acceleration technologies to shorten data processing time and meet higher real-time requirements. In addition, strengthening the design of the user interface, improving the convenience of operation and the usability of the system will also be an important direction for future research.

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