

UNDERWATER IMAGE CLASSIFICATION AND ENHANCEMENT

¹Miss. Swati Sonawane, ²Prof. Dr. R.N. Awale

¹M. Tech (Electronics & Telecommunication) 2022, ²Professor in Electrical Engineering Department
VJTI, Mumbai, India

Abstract: Image Enhancement is an important step in underwater research. The basic intent of image enhancing is to convert a blur image into a crystal-clear image for the underwater research. So, this paper discusses the technique for improving underwater image enhancement, these underwater images usually suffer from motion blur effect due to turbulence in the flow of water and non – uniform illumination and limited contrast. Due to the presence of distortion, captured underwater image needs to be processed in different ways. Underwater images captured in deep low light environment, are of worst quality and these images are low contrast, cause blurring effect, low contrast, scattering, absorption, noise color variation, clarity of image is reduced, quality get degrades and these underwater images cannot be directly used for various scientific research, marine biology research, underwater vehicles, submarine operations. While capturing underwater images some major obstacles are there such as minerals, salt, sand, planktons. These particles produce haziness in deep underwater captured image. To beat this, transfer learning base of Features model is taken on in this paper.

Keywords: Underwater Image, Image Enhancement, Transfer Learning, Light Scattering

INTRODUCTION

The Underwater image enhancement techniques is used, because the earth is planet having 70% of its surface is covered by water, and underwater imaging has vast application as the river, sea, lakes, and oceans contain many valuable resources inside them, So, scientists and researchers have shown great interest in capturing underwater life. It is observed that the effect of scattering and absorption of light in water are the major causes of it. When light enters from air to water it suffers dispersion, scattering effect, when it strikes particles of sand and minerals dissolved in water. Scattering deflects light in different directions reducing the amount of light falling on the object captured

Underwater images have also been an important source of interest in various branches of technology and scientific research. These techniques are widely used in numerous applications, such as the inspection of submarine infrastructure and cables.

Image enhancement is to bring more visibility to the image and make it more appropriate to the required application. In today's scenario, the process of underwater image enhancement becomes an important area of study. The quality of underwater images deteriorates due to the physical properties of the aquatic medium, light scattering, reflection, and becomes more and less visible as water depth increases. The haziness is caused by suspended particles such as sand, minerals, and plankton that exist in lakes, oceans, rivers, sea. As the light reflected from the objects advances towards the camera, a part of the light meets these suspended particles, which absorb and disperse the light. Capturing clear images underwater is a challenge, mainly due to the turbidity caused by the dispersion of the color, in addition to the color emitted by the attenuation of the variable light at different wavelengths. Color dispersion and color emission produce blurred subjects and low-intensity contrast in underwater images.

LITERATURE SURVEY

Enhance the underwater images that are degraded because of the scattering, absorption of the medium.

single image method for underwater images which calculate the white balance and then two variants of the image made one for which the correction is being calculated and other for which the sharpening is calculated from the resultant image which is white balanced then the weight-maps are being applied and finally multi-fusion technique is applied for getting the final result their approach is able to improve many varieties of images captured under the water with accuracy.

Multi-scale Fusion technique calculated for Laplacian pyramid guided by the weight maps Number of pyramids increase with the image size, they introduce multi-scale fusion based on Laplacian decomposition. The underwater environments suffer from dispersion and absorption phenomena that disturb the visualization of the image and propagation of light, degrading the quality of underwater images. A physical model of light propagation method and the use of previous statistical data can restore the image quality achieved in the typical underwater scene.

Underwater exploration has increased in recent years exponentially. Equipment currently available for data collection (side scan sonar, multi-beam sonar, sub-bottom profiler, remotely operated vehicle)

Underwater research and observations not only provide data on objects and species. It also provides data on sea level. For this purpose, the selection of suitable characteristics is hard work. Classification is difficult due to limited underwater datasets Objects/features from underwater images. To overcome this, machine learning. A bag-of-features model is used in this document. Click here for the record Shoal with ROV. Because there is little light in optical underwater images, the strength that makes feature classification a difficult task. SURF (acceleration Robust Features) and SVM (Support Vector Machines) algorithms are implemented. Achieve maximum accuracy with the Bag of Features model. Performance evaluation Combining training and testing datasets improves performance.

Considered as object-based image analysis (OBIA).

It is an effective technique for high spatial resolution (HSR) imaging. Classification by a clear and intuitive technical process. However, OBIA relies on manual adjustment of the image. Classification function. This is tricky work. Deep learning (DL) The technology automatically learns image features from a large number of images, Achieving higher image classification accuracy than before Technique. The study uses a new method called object scale adaptive convolutional neural networks (OSA-CNN), Combine OBIA and CNN, recommended for HSR images classification. First, OSA-CNN collects image blobs

Principal axis of the object primitive taken from the image segmentation; the size of the former is determined automatically By the axial width of the latter. This step generates the input Units required for CNN classification. Second The squeeze and excitation blocks are extracted from the SE network.

The network structure of Google Net that realizes this

Improved weighted merging for multiscale convolution functions Suppress useful functions and suppress useless functions. when classifying stadium, multiscale image segmentation, CNN classification

It is fused using the object scale adaptation mechanism. contradiction at the end

Primitives are classified by majority vote over the image Dirt. Changes in network structure, multiscale

Classification fusion and other improvements

Gradually integrate these steps into the original Google Net. Trials Show These Improvements Are Effective Improved image classification accuracy. This research an effective way to leverage a combination of OBIA and DL techniques Advantages of both approaches and promotion of HSR image classification.

An Accurate and Robust Classification Method for Sea Ice and Sea Ice Open water is important for many applications. Synthesis Aperture Radar (SAR) Imaging Capabilities A meteorological condition, often used to classify sea ice. U-Net, a deep learning framework, is doing great work Success in the field of biomedical image classification. In this

The study builds a U-Net-based 'end-to-end' model.

Classify sea ice and open water pixels in SAR images.

Five SAR images taken in the Gulf of Alaska near Bering A strait is used in this case study. Manually label the SAR an image of ice and water. Images labeled from scratch Four SAR images are split into chips and fed to the U-Net model for training. The fifth SAR image is used as follows. test data. The experiment is for accuracy and The test pattern recall is 91.64% and 91.70%, Respectively. Most sea ice containing small chunks and A winding ice edge can be successfully classified.

Remote sensing has been stood out in areas such as agriculture, disaster control, water resources management, urban planning among others. The process of generating a thematic map from these satellite captured images known as image classification. The aim of this work is to develop a software that can use many spectral bands from the satellite payload in order to classify it and detect water, using supervised or not supervised learning techniques. So that result can to be apply in water bodies detection. The information obtained from the analysis could be used in protected natural areas, for instance “Cuatro Ciénegas” which is known as endemic area, this kind of regions needs monitoring their behavior over time.

SYSTEM ARCHITECTURE

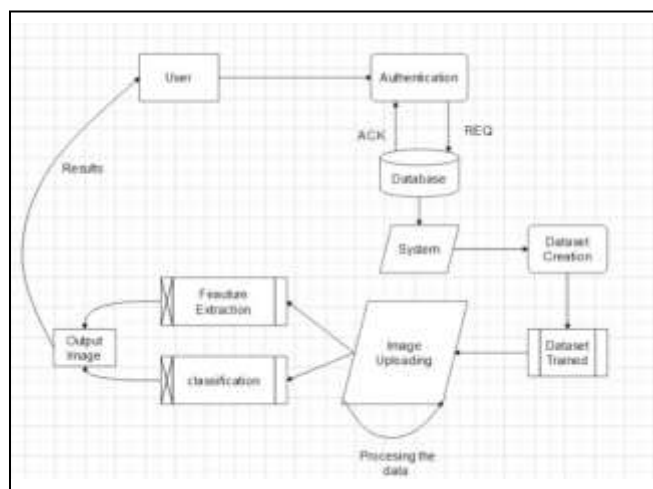


Fig. System Architecture

In the above fig, we have mentioned the work of our system where firstly will upload the image to system, in that process the system will take input and start the process of filtering, where extra noise and header is removed and formed a regression, system save it.

The system then extraction process is carried out same as previous like noise and frame header removal. After that, the result is stored separately.

Then the two results are combined and the system matches the difference and provides output.

ALGORITHM

Step 1: Start

Step 2: Log In with your User ID and Password. Step 3: Upload the Blur Underwater Image.

Step 4: System Use Transfer Learning Algorithm of Machine Learning.

Step 5: With the help of the system, the Blur Image is Converted into a Clear Image and it will classify the image in output

Step 6: The output should be displayed on the screen. Step 7: Stop.

EXPERIMENTAL RESULTS



Fig. Image Uploading Screen



Fig. Screen after Uploading Blur Image



Fig. Clear Image Output

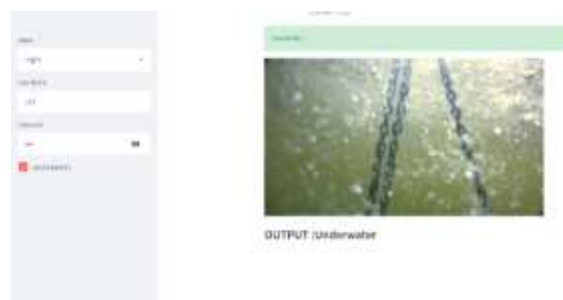
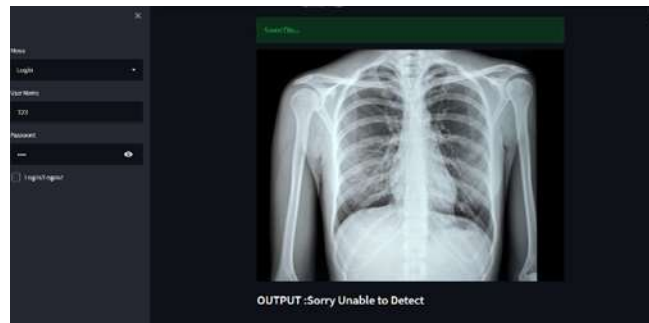


Fig. shows the output as it is underwater image



Fig, shows output as unable to detect.

CONCLUSION

In this paper the problem related to Underwater Image Enhancement for finding crystal clear underwater images is a great challenge, and the presence of scattering and absorption in underwater pictures create difficulties, we have examined a technique for enhancement which have been specifically developed for the underwater pictures, and we will find results from the output image. these methods work on all the underwater images, which eliminate obstacles and develops a simpler and more effective image. Similarly, we have make the sorting of this images in the output.

REFERENCES

- [1] A. S. Bhadouria, "Effective Preprocessing for deep underwater image enhancement," Springer conference,
- [2] D. Bhadoriya, "A Block Based Scheme using Tuned Tri-threshold Fuzzy Intensification Operators for Underwater Images," 2019. (Fuzzy)
- [3] O. Ancuti, C. Ancuti, "Color balance and fusion for underwater image enhancement," IEEE Transaction on Image Processing, Vol. 27, NO. 1, January 2018
- [4] Pan-wang Pan, Fei Yuan, "Underwater Image DE-Scattering and Enhancing using Dehazenet and HWD, Journal of Marine Science and Technology[J], Vol. 26, No. 4, pp. 531-540 (2018).
- [5] K. He, J. Sun and X. Tang, "Single Image Haze Removal Using Dark Channel Prior," IEEE Transactions on Pattern Analysis and Machine Intelligence, Volume 30, December 2011. (DCP)
- [6] G. Yadav, S. Maheshwari, "Contrast limited adaptive histogram equalization-based enhancement for real time video system," IEEE Xplore, 01, December 2014. (CLAHE)
- [7] R. Fattal, "Single image dehazing," in Proc. ACM SIGGRAPH, Aug. 2008, Art. no. 72.
- [8] Aashi Singh Bhadouria, "Underwater Image Enhancement Techniques: An Exhaustive Study" IJRASET, Volume 10, Feb 2022
- [9] M. Vimal Raj and S. Sakthivel Murugan, "Underwater Image Classification using Machine Learning Technique", IEEE PROCEEDINGS OF SYMPOL-2019
- [10] Jie Wang, Yalan Zheng, Min Wang, Qian Shen, and Jiru Huang, "Object-scale adaptive convolutional neural networks for high-spatial resolution remote sensing image classification", IEEE Journal 2020