

# Underwater Image Enhancement Using the Multilayer Perceptron and Color Feature-Based SRCNN

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# ABSTRACT

- ▶ **Problems in Traditional Approach** : The traditional underwater visual display systems have problems, such as low visibility, poor real-time performance, and low resolution, and cannot meet the needs of real-time high definition displays in extreme environments.
- ▶ **Solution** : An underwater image enhancement method and a corresponding image super-resolution algorithm.
- ▶ Image enhancement Technique.
- ▶ Super-resolution Convolutional neural networks
- ▶ the Retinex algorithm
- ▶ gamma correction.
- ▶ Dark prior

# SRCNN MODEL

- ▶ The SRCNN network structure is constructed by a three-layer convolutional neural network, which is composed of image block feature extraction and representation, and also nonlinear mapping and reconstruction of high-resolution images.
- ▶ First layer, feature extraction and representation is applied to the image blocks from the original image, and the image blocks extracted by convolution can be expressed as

$$F1(Y) = \max(0, W1 * Y + B1) \quad \text{----(1)}$$

- ▶ where  $Y$  represents the original high-resolution image,  $W1$  and  $B1$  represent the convolution kernel and deviation.
- ▶ Second layer of nonlinear mapping is as follows:

$$F2(Y) = \max(0, W2 * F1(Y) + B2) \quad \text{-----(2)}$$

- ▶ Third layer, a convolution is conducted on the output high resolution image blocks of the second layer to generate images that are close to the original high-resolution images, which can be expressed as:

$$F3(Y) = \max(0, W3 * F2(Y) + B3) \quad \text{-----(3)}$$

# PROPOSED METHOD

## MULTILAYER PERCEPTRON-BASED ENHANCEMENT

- ▶ First, Retinex algorithm is used to initially defog the image. Second, due to the low contrast of underwater images, the image brightness was adjusted by gamma correction to make the image more natural.
- ▶ The preprocessed image can be obtained from the following equation:

$$r'(x, y) = r''(x, y)$$

- ▶ where  $r'(x, y)$  is the image enhanced with the Retinex algorithm and  $r''(x, y)$  is the Gamma corrected map.
- ▶ Finally, a dark channel prior is used, and the contrast stretch technique is applied to improve the dynamic range of the image.

# Output of MULTILAYER PERCEPTRON-BASED ENHANCEMENT

Original Image



After Retinex



After Dark Prior



After Gamma Correction





# COLOR FEATURE-BASED SUPERRESOLUTION

- ▶ The method in this paper is based on an SRCNN composed of three convolution layers to improve the input image. The method in this paper is divided into three steps.
- ▶ The first step is to divide the low-resolution image into three separate images with three RGB channels.
- ▶ In the second step, the three images are trained by a convolutional neural network to obtain three output images. In the
- ▶ Third step, the obtained three images are fused to obtain the high-resolution image reconstruction.
- ▶ The first step is channel processing. The original image is divided into R, G and B channels to extract different information from each channel. The segmentation processing formula can be expressed as:

$$F_i = \text{image}(F_i) \quad (i = R, G, B)$$

- ▶ The second step is the convolutional neural network training. The images of the three different channels are trained by a convolutional neural network. The training formula can be expressed as:

$$F_i = \max(0, W_j * F_i(j-1) + B_j) \quad (j = 1, 2, 3)$$

- ▶ The third step is image fusion, where the output image of the second step is fused. The fusion formula can be expressed as:

$$F = \text{Image}(F_r, F_g, F_b)$$

- ▶ Dataset used : **UIEB Dataset Size (630 MB)** 890 raw underwater images with corresponding high-quality reference images
- ▶ **SRCNN Training**
- ▶ **Training SRCNN 100 Epochs**
- ▶ **Adam Optimizer (0.001 learning rate)**
- ▶ **Testing on** 60 challenging underwater images.

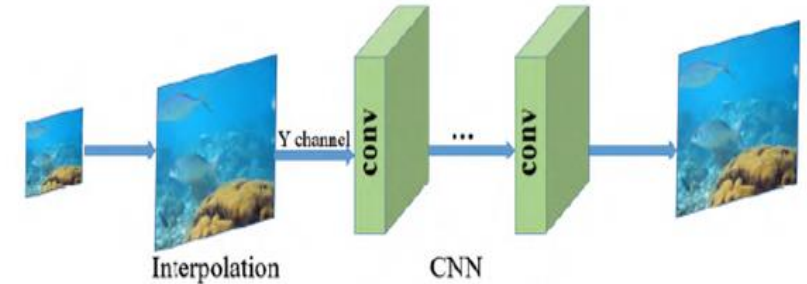


FIGURE 1. Flowchart of the SRCNN.

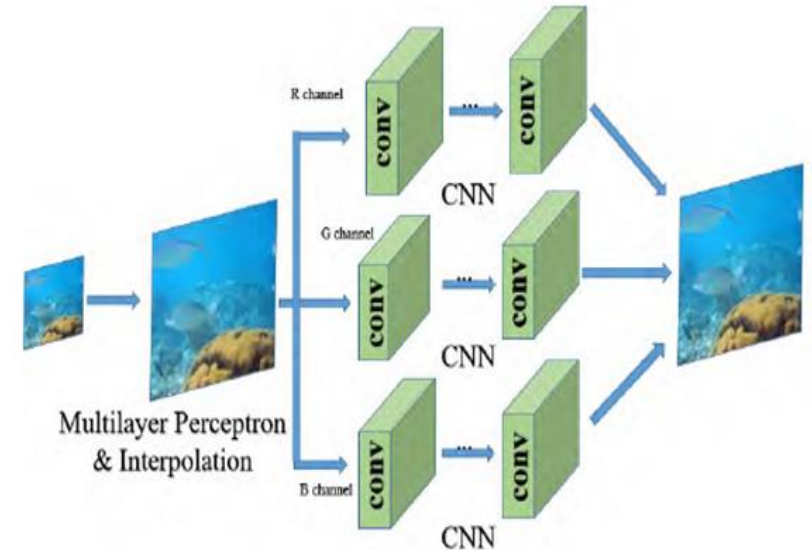


FIGURE 2. Flowchart of the proposed method.

# EXPERIMENTAL RESULTS AND DISCUSSION

## ► VISUAL ANALYSIS



Original image



Retinex image



Gamma Image



Dark Prior



### SRCNN Images



Red channel

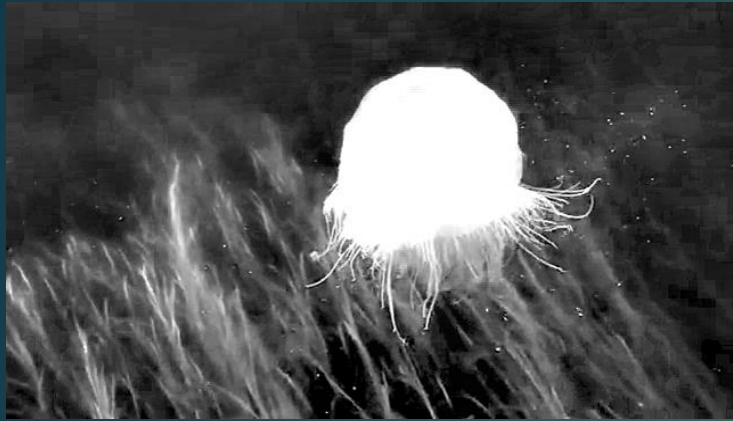


Blue channel



Green channel





Red Channel



Blue Channel



Green Channel



Fused Image/Output Image

# IMAGE QUALITY EVALUATION

**PSNR** : peak signal-to-noise ratio, ii between two images.

This ratio is used as a quality measurement between the original and a compressed image.

Image Channels	PSNR value
Red channel	29.4263
Green channel	28.6954
Blue channel	31.6093

Images no.	PSNR value	SSIM Value	PSNR/SSIM
image1	15.135387224955284	0.5460245819947694	27.719241448181304

# Conclusion

- ▶ This paper presented an improved image enhancement method and an image super resolution method. The proposed enhancement method employed a combination of the Retinex algorithm and a neural network to enhance the details of the image and restore the image color.
- ▶ Referring to the network structure of an SRCNN, a convolutional neural network model that is trained on the R, G and B channels of an image is adopted, and the output images are fused to obtain clear textures and edge effects under the premise of ensuring the PSNR indexes.
- ▶ a good reconstruction effect on most images, its advantages are not obvious for images with nonobvious edges and irregular textures.
- ▶ The next step is to try to classify and reconstruct the edges and textures to generate images with clear edges and rich textures and further study image fusion.
- ▶ I will further try the **SRGAN Model** to reconstruct the edges and textures to generate images with clear edges and rich textures



# Thank You

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