



HISTOGRAM BASED RESOLUTION OF AN IMAGE BY USING CONVOLUTIONAL NEURAL NETWORK

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Abstract

Compared to low resolution (LR) photos, high resolution (HR) photographs include more information about the image [1]. The HR picture may be retrieved from many low resolution images more readily than it can from a single LR image. To create an HR picture, many LR photos that contain different features can be combined. It is very difficult to reproduce the HR picture with a single, less subtle LR image. A neural network based model which is feature based extraction from the single picture is suggested in this research to improve the resolution. Every LR and HR picture is broken up into tiny blocks and each block in the histogram is estimated by a new feature. The histogram properties of LR are learned using back propagation neural network (BPNN). The simulation is carried out on a set of MRI images of the brain, and the findings state that the PSNR and RSME has improved by the resultant neural network.

Keywords: Histogram, resolution, Digital image processing, Neural network

Introduction

The quality of the visual information affects the effectiveness of applications including imaging the medical based applications, remote sensing video surveillance and conferences. To improve the validity of these systems, HR photos are necessary. If the HR picture is provided, segmentation of the region and the efficiency of the recognition of pattern in computer vision will be improved when using HR satellite images.

For these effects the alternate Back-Projection approach was developed, however due to the oppositeness of the inverse issue, it is unable to offer a single solution.

The resolution of medical photographs has been improved using a learning-based algorithm.

A variety of deep learning techniques were used to improve image resolution, ranging from the early Convolutional Neural Networks (CNN) based approach to super-resolution [2].

Techniques for improvement are based on Generative Adversarial Nets (GAN). In this study, we proposed a learning-based technique for increasing the resolution of brain MRI images.

Images, a crucial multi media information medium, are deeply ingrained in our daily life and convey a wealth of information. Computer vision based applications like image target categorization and target identification were continuously developing thanks to the quick development of technology of Deep Learning, which makes life easier for people.

In order to do this, the computer must extract the primary scene from the flawed scene. In order to generate HR images and use them for feature extraction, super-resolution (SR) [3] based techniques upscale low-resolution (LR) images.

These photos that need to be restored typically have a variety of problems, such as noise or the information shortage. Images are the primary information collected by the vision system, but the low quality images that are required to be used do not include the information that the system requires, reducing the efficiency and computer vision system's accuracy.

Studying the method for enhancement processing on degraded pictures is crucial for this reason. They were incorporated into several models of Deep learning after model selection. Calculate the average of each of their individual forecast findings. The impact is better the more the models differ from one another. It is useful in overcoming the algorithm's generalization ability [1].

Currently, high visibility images and images with high feature density are used in recognition of images or other tests in computer vision and the computer vision tasks are performed with these images. Features and aesthetics have produced exciting results [2].

If each model is autonomous and successful using quite diverse network topologies and technologies, the consequence of the combined outcome is more solid. On the other hand, you can run experiments sequentially. The network model is initialised and the final weight will meet the clear value each time while it was trained. To overcome the algorithm's propensity to generalise, this procedure is done several times to produce different models of network and then the estimated outcomes of the models are integrated.

The computer vision based tasks are performed under these images with obvious characteristics. In accordance with aesthetics we produced intriguing results. High image visibility, image clearance and feature density images were used in image recognition of images and other tests in computer vision.

Image enhancement & reconstruction applications include smoothing the image and denoising, deblurring the image, rain and snow, image super resolution, picture inpainting, detailed improvement & mapping of high dynamic range tone.

Both parties want to develop the visible image standard & in some ways, receive an enhanced image. Image enhancement, on the other hand, does not seek to restore the image's original appearance; rather, it selectively emphasizes some interesting details and useful information while muting or weakening others, producing results that are expressive, obvious in their characteristics, and rich in useful information.

with an increased number of iterations, it can even produce outcomes that are comparable to those of the most advanced convolutional network-based systems converge on VDSR [5].

Depth neural network mapping association is examined in this study. An estimation function for the illumination map is presented. For local adjustment of the brightness information of the source data in order to preserve the photo graphs' high dynamic range in low lighting conditions, the illumination map is utilised. Transmittance of the media is calculated with the neural network is used in this study to improve the details of the augmented image [1].

It also mentions the variable of detailed development methods which have a significant effect on image enhancement outcome. The paper employs a multi objective optimisation technique to deal with these super parameters to attain steady and operative results. Appropriately to deal this problem, this system employs the differential evolution algorithm.

Proposed Method

In the proposed method, a single LR picture is used to boost an image's resolution by two times

Parameter	Proposed Method	Existing Method
PSNR	16.9538	5.2348
RMSE	0.0202	0.2988

Table: Comparison of PSNR and RMSE Simulation Results

In order to produce a couple of picture blocks which are low & corresponding high dimensional block, the histogram was employed as a feature. This technique is applied to a dataset of 31 LR and associated HR brain MRI pictures. LR images are [128 by 128] in size, whereas HR images are [256 x 256]. The training of neural network used in the suggested approach was based on error back propagation.

The following flow chart represents the proposed work :

Flow chart for proposed work

Each image was separated into blocks before the histogram feature was extracted. The size of each block for LR and HR photos was $[2 \times 2]$ and $[4 \times 4]$ respectively, which results 4096 blocks in each every image. The neural network was trained using 21 brain MRI scans [6], resulting in a total $[21 \times 4096]$ blocks for both HR and LR picture.

The Low Resolution image set and the corresponding High Resolution image set are referred to as super resolution brain MRI images. According to the results, the developed model increases the PSNR & RMSE value.

The saved LR with the ID of the class replaces the relevant Low Resolution block. The associated LR block is then substituted by this one. By extracting the histogram feature from a single LR image, the HR block generates an HR image.

Simulation Results

The simulation of the developed algorithm is executed and operated on the MATLAB

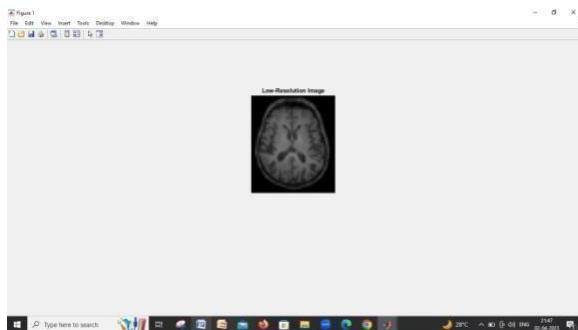


Fig1:Low Resolution Image

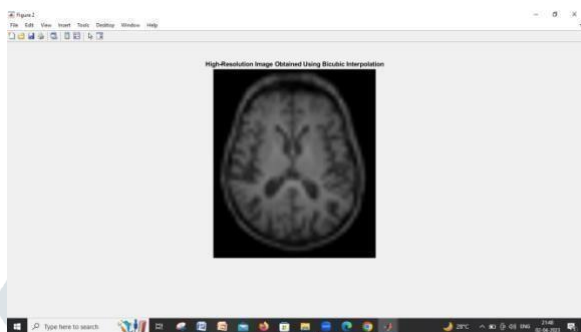


Fig2:High Resolution Image attained by using Bicubic Interpolation.

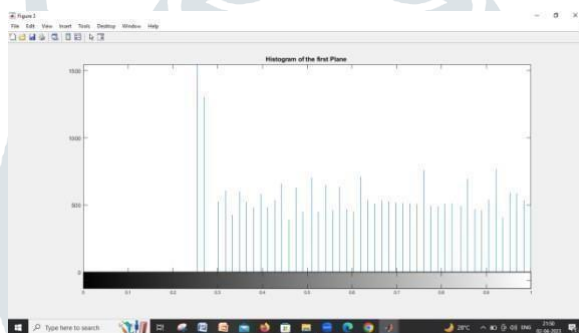


Fig3:Histogram of the first plane.

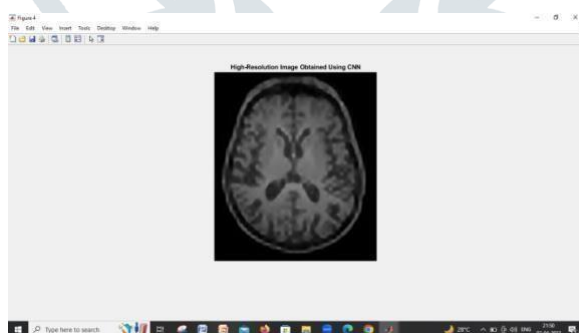


Fig4: High Resolution Image obtained by using CNN.

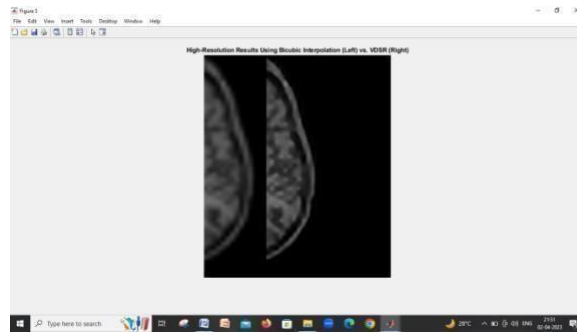


Fig5:High Resolution Results by using Bicubic Interpolation (left) vs VDSR(right).

Conclusion:

By developing a neural network for image transformation from LR to HR, this paper has improved the work of image resolution enhancement. A histogram feature with 16bins was extracted to increase the resolution of a single LR image. This histogram feature was used to train a back propagation neural network model for the selection of appropriate class ids. The simulations were performed on a collection of brain MRI images.

In order to reduce the experiment's external validity threat, more datasets and image enhancement algorithms will be used in future work. Further more, the image enhancement algorithms will be chosen based on the dataset characteristics. Further more, using more CNN models and transfer learning strategies can reduce external effectiveness threats in experiments. Further more, more experiments can be designed to investigate the effect of Laplace operators on the performance of CNN models.

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