Resolution Enhancement Using Frequency Domain Interpolation

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Abstract—In various fields of digital signal processing and image analysis, the need to enhance the resolution of low-quality or under sampled data is a common challenge. This abstract introduces a novel approach to resolution enhancement, utilizing Frequency Domain Interpolation (FDI) techniques. Resolution enhancement is a fundamental requirement in applications such as medical imaging, remote sensing, surveillance, and computer vision. Traditional interpolation methods, like bilinear or bicubic, often lead to undesirable artifacts and limited improvements in image quality, particularly when significant upscaling is required. The proposed approach leverages the frequency domain as an alternative to spatial domain techniques, offering advantages such as reduced aliasing and superior preservation of image details. By exploiting the rich spectral information present in the input data, FDI aims to provide more accurate and visually pleasing high-resolution reconstructions. We do the experiment on 128x128,512x512 images using nearest neighbour, bicubic, bilinear interpolation and wavelet method.

Key words: Frequency Domain Interpolation

I. INTRODUCTION

Spatial domain refers to the normal image space represented as a matrix of pixels. Transformation techniques in this domain are directly operated on image pixel values. The values are manipulated to achieve desired enhancement. Frequency domain deals with the rate at which these pixel values change in spatial domain. Frequency simply refers to the rate of change of color components in an image. Areas of high frequencies experience rapid color changes, whereas parts that change gradually contain low frequencies. Unlike spatial domain, we cannot directly operate on the values. The image is first transformed to its frequency distribution before processing it. These frequency components are divided into two major components. High Frequency components that correspond to edges in an image and low frequency component that correspond to smooth regions. The output of this process isn't an image, rather a transformation. To reconstruct the image to its ideal form, we need to apply inverse transformation on the processed output. Although there are several mathematical transforms under frequency domain such as the Fourier transform, Laplace transform, Z transform, this article will explore the wavelet transformation technique that is generally used for image analysis. compression. Wavelet Transformational wavelet is a wave-like oscillation with an amplitude that begins at zero, increases, and then decreases back to zero. It can typically be visualized as a "brief oscillation" like one recorded by a seismograph or heart monitor. Wavelets are functions that are concentrated in time and frequency around a certain point. This transformation technique is used to overcome the drawbacks of Fourier method. Fourier transformation, although it deals with frequencies, does not provide temporal details. According to Heisenberg's Uncertainty Principle, we can either have high frequency resolution and poor temporal resolution or vice versa. This wavelet transform finds its most appropriate use in non-stationary signals. This transformation achieves good frequency resolution for low-frequency components and high temporal resolution for high-frequency components. This method starts with a mother wavelet such as Haar, Morlet, Daubechies, etc. The signal is then essentially translated into scaled and shifted versions of mother wavelet.

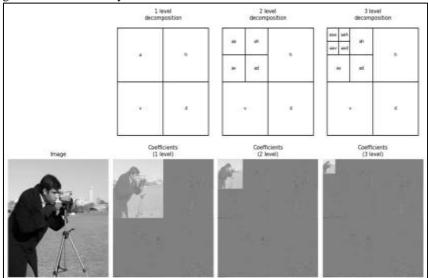


Fig. 1: Visual comparison of different interpolation methods. Wavelet analysis is used to divide information present on an image (signals) into two discrete components — approximations and details (sub-signals). (a) Nearest-neighbor. (b) Bicubic. (c) wavelet (d) Original HR image .image1

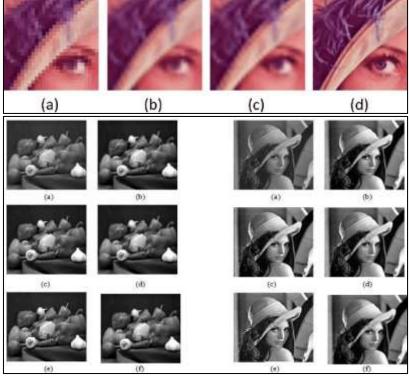


Fig. 2:

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	PSNR	MSE	
Nearest neighbor	29.05	.0134	
bilinear	31.32	.0132	
bicubic	32.33	.0034	
Proposed wavelet	34.38	.0012	

Table 1: RESULT PSNR and MSE (Image1)

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Method	PSNR	MSE	
Bilinear	23.33	.0145	
Bicubic	25.80	.0142	
Nearest Neighbour	26.04	0.0032	
ProposedMethod	26.96	0.0010	

Table 2: RESULT PSNR and MSE (Image2)

II. CONCLUSION

The application of Frequency Domain Interpolation (FDI) to resolution enhancement represents a significant advancement in the field of digital imagery. Through this research, we have explored the potential of FDI to address the long-standing challenges associated with traditional interpolation methods and have presented compelling evidence of its effectiveness. Frequency Domain Interpolation has emerged as a compelling and innovative technique for resolution enhancement, with the potential to revolutionize image processing and analysis across numerous disciplines. Its ability to enhance the quality and detail of images in a manner superior to traditional interpolation methods signifies a significant leap forward. As researchers continue to explore and refine this approach, the future holds the promise of even more precise and impactful resolution enhancement techniques that will benefit society through improved decision-making, diagnostics, and analysis across a myriad of applications, the future scope of resolution enhancement using Frequency Domain Interpolation is extensive and holds the potential to transform multiple industries. As technology advances and research continues, FDI will likely play a pivotal role in improving the quality and accuracy of digital imagery, benefiting fields as diverse as healthcare, environmental monitoring, security, and entertainment. Research and innovation in these directions will shape the future landscape of image resolution.

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