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¹ Exerting Moment Algorithms for Restoration of Blurred Images

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6 Abstract

In this paper presents the restoration of blurred images which gets degraded due to diverse 7 atmospheric and environmental conditions, so it is essential to restore the original image. The 8 research outcomes exhibit the major identified bottleneck for restoration is to deal with the 9 blurred image as an input to imaging agent employing various methodologies ranging from 10 principle component analysis to momentary algorithms and also a set of attempts are been 11 executed in image restoration using various algorithms. However the precise results are not 12 been proposed and demonstrated in the comparable researches. Also detail understanding for 13 applications of moment algorithms for image restoration and demonstrating the benefits of 14 geometric and orthogonal moments are becoming the recent requirements for research. 15

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17 Index terms— image descriptors, moment algorithm, image blurring, legendre moment, image restoration.

18 1 INTRODUCTION

mage processing is a very active research area that has impact in several fields from remote sensing, Biometric 19 authentication system, robotics, traffic Surveillance, to medicine. Automatic target recognition and Tracking, 20 character recognition, 3-D scene analysis and reconstruction are only a few objectives to deal with. Since the 21 real sensing systems are sometimes imperfect and also the environmental conditions are dynamic over time, the 22 acquired images often The image are the for the most part frequent component of information representation 23 and transmission due to the robust nature of information storage and the continuous effort to make digital 24 25 image processing and presentation better. The studies have shown that the images contain information which is 26 redundant and changing a value may cause errors in the calculation for further steps.

In the space of image processing, the restoration of images is the major expanse of research Author? ?: Depart-27 ment of CSE BVRIT Narsapur Hyderabad, India. e-mails: bhima.mnnit@gmail.com, jagan.amgoth@bvrit.ac.in 28 for many decades. Many researchers have proposed various algorithms and techniques for better restoration of 29 images for various applications. However the collection of image is strongly dependent on the imaging agent. 30 The quality of a image possibly will suffer from a variety of impairments, Still the key bottleneck for better 31 restoration of images are the random distortion and blurring caused to the initial images to be provided as input 32 to the recognition system [1] [2]. The distortion and blurriness of the images are not only dependent on the 33 capture agent, but also depends on the environmental and human errors. The causes of blurriness are studies 34 and classified in four major kinds. Firstly, the focal length of the capture devices, Secondly, during the capture 35 36 of object in a time irrelevant scale needs to be mapped with the capture speed of the agent to avoid the blurriness 37 [3]. Thirdly, sometimes due to environmental and human causes the stabilization of the capture devices may 38 be disturbed causing the blurriness. Fourthly, the most unavoidable situation, where the object is in higher order of colour range but the relevant background of lower order of colour range causing the blurriness. Thus to 39 remove the effect of blurriness of the image, the most appropriate algorithms to be deployed are the momentary 40 calculation algorithms. 41

In the field of image processing, computer vision and allied fields, an image moment is a certain particular weighted average (moment) of the image pixels' intensities, or a function of such moments, usually chosen to have some attractive property or interpretation. Image moments are helpful to depict objects after segmentation.

6 APPLICABILITY OF LEGENDRE MOMENT

45 Simple properties of the image which are found via image moments include area (or total intensity), it's centric,

⁴⁶ and information about its orientation and Effects of moments in digital image processing for restoration cannot ⁴⁷ be ignored as supported by related researches. In general moments are the numeric values used to represent

the nature of any functions and identify with the significant properties ??3] [4]. The following are mostly used moments algorithms are Hu moment, Zernike moment and the well discussed Legendre algorithms.

50 The moments are superior to principle component analysis for image recognition especially for image

recognition [5] [6] [7]. Yet the application of moments algorithms are not been studied for digital image restoration

⁵² with the comparative results for blur to 7 Year 2016 () restoration algorithm efficiency mapping. Thus in this ⁵³ work we understand the algorithms of moments calculation proposed by Hu, Zernike and Legendre for image

⁵⁴ restoration and develop a framework for comparing the visual performance of the restoration process by applying

55 the same algorithms.

This work also demonstrates the effect of multi order Legendre for blurred image restoration. The rest of the work is organized as in Section II we understand the basic constructions of the moment algorithms and possibilities to apply for image restoration, in Section III we consider the Legendre moment in detail, in Section

59 IV we define the components for Blurred image restoration process, in Section V, we demonstrate the proposed 60 framework for Blurred Image recovery using multi-order Legendre moment algorithm, in Section VI we discuss

the application constructed for the visual comparison for the blurred image restoration, in Section VII we discuss

62 the results tested on multiple image datasets and in Section VIII we discuss the conclusions and future scope of 63 this work.

⁶⁴ **2 II.**

65 3 IMAGE MOMENTS

In Image processing and computer vision processing explore the calculation of image moments or finding the image descriptors is widely accepted. The moment is a calculated on certain weighted average of any pixel taking into account the neighbourhood pixel values. Often the moment is also used to calculate to understand and extract the most significant property of a continuous function [8]. The image moments are widely accepted for image processing and used by all polynomial approaches. In this work we consider the restoration techniques

⁷⁰ image processing and used by an polynomial approaches. In this work we consider the restoration technic ⁷¹ using moments, thus the understanding of moments will be helpful in section IV.

In case of image and vision processing calculating the image moment which is resulting in the image descriptor is performed after the image segmentation. The image properties like area, centroid, pixel values and orientation of any object in the image can be represented using the image moment.

Image moments are classified into three categories such as Raw Moments, Central Moments and Scale invariant
 Moments [3]. In this work, we understand the moments in details:a) Raw Moment

For a simple two dimensional function, denoted by (,) f a b , the raw moment of order (,)

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82 4 .().().

$_{s5}$ 5 c) Scale invariant Moment

The moment of order (x + y) where x + y? 2 can be obtained by dividing the central moment with 0 th moment as following:

6 APPLICABILITY OF LEGENDRE MOMENT

The most adopted method for image pattern or image restoration is the use of moments. The recent advancements demonstrate the use of moment calculation methodologies as geometric and orthogonal moments. Further studies have demonstrated that the orthogonal moments are better than the geometric moments. Among the orthogonal moments the most widely accepted method is to deploy the Legendre moment. But the application of Legendre moment is also restricted for the blurred or distorted images. Here we understand Legendre Moments in detail [3] [13]: Legendre Moment for of order (a + b) is defined as:1 1 1 1 (2 1)(2 1) (). (,). 4 ab a b a b P i P i j didj ? + +?? + + =??

Where a, b is ranging from 1 to?. Hence the k th order Legendre polynomial is written as: 2 2 (2)! (2)! () 97 ... 2 (!) 2 !(1)!(2)! k k k k k k k P i i i k k k? ? = ? + ? ? K th Term ?Eq9 Where, D(k) = k/2 or (k-1)/2,

98 is an positive integer.

99 IV.

7 CHARACTERISTICS OF BLURRED IMAGE

In the Blurred or noisy image, the objects vary in terms of contrast and size. The objects in the image can represent large to small item or the items with detailed visibility. The primary effect of the blurriness on the imageis to reduce the contrast and visibility of the images. The reduced visibility images causes less detailed information in the images [10] [11].

The objects in the images are generally differentiated by the pixel difference between the object and the background at the object edges. The blurriness of the image actually reduces the pixel difference at the object edges [12].

The blurriness of the image can be measured in terms of units of lengths. The length of the images denotes the blurriness of the image [Table ??

¹¹⁰ 8 FRAMEWORK FOR BLURRED IMAGE RESTORATION ¹¹¹ PROCESS

112 The two dimensional Legendre Moment for the blurred image of g (a, b) can be defined as [3] [13]:11, 11

¹¹³ 9 () (). (). (,).

114 x y x y L g P a P b g a b dadb+ + ? ? = ? ? ?Eq10

With the understanding of blurriness effect on the image, the image pixel will be multiplied by random value generated by the noise function. 11, 11

- 117 ()().().(*).
- 118 x y x y L g P a P b f h dadb+ + ? ? = ? ? ?Eq 11
- Legendre moment of the blurred image can be represented as 1 1, 11

122 x y x y L g h i j P a i P b j f a b dadb didj+ + +? +?????? = + +????? (P a f a b dadb didj+ + +??????? = + +??????

123 Image restoration procedure using moments:-

- 124 ? Capturing image using capture device.
- ? Captured image is stored and referred for pre processing ? Blur function is applied on Image and also calculates image moment using Legendre polynomials.
- ¹²⁶ Comparison of original blurred and restored image.

Thus the process of restoring the blurred image using Legendre Moment is presented in this work [Figure -1]. visual advantages of Legendre Moments over other available moments. MATLAB is a highly popular multipurpose numeric programming language for the wide variety of build in library functions ranging from image processing to higher order numeric calculation. The built in library is capable of generating matrix based calculation and graph plotting in multi-dimensional space. The MATLAB is considered as the fourth generation programming language.

In the implementation we also propose the multi order Legendre Moments to restore the blurred and noisy image.

136 **11 VII.**

137 12 RESULTS AND DISCUSSIONS

In this section, we have considered three different image dataset of fingerprint, bird and human face for restoration using various methods such as Hu, Zernike and Legendre moments. Henceforth we compare the initial image and restored image generated by the Hu, Zernike and Legendre moments using the following formulation:

141 The difference between the original image and the restored image using movements' algorithms considered as 142 K 1 and the difference between the original image and blurred image is considered as K 2.

Hence the comparative difference between the K 1 and K 2 is considered K, demonstrating the amount of successful restoration for any given image using any given moment algorithm. The testing results clearly demonstrate the comparative study on different data sets such as fingerprint, bird and human face for restoration using Hu, Zernike and Legendre moment. For fingerprint Hue method exhibit better results, Zernike and Legendre

147 shows better results for bird. In the case of human face Legendre moments demonstrates better results.

¹⁴⁸ **13 VIII.**

149 14 Conclusion

150 In this paper focus on the analysis of three categories of moments such as Raw Moments, Central Moments

and Scale invariant Moments and the basic mathematics functions behind those moments. In order to achieve better understanding of image restoration process, we have also understood the nature of blurred images. The understanding of the difference of lengths for normal and blurred image based on the length for various capture

14 CONCLUSION

device types. Henceforth, this work proposes a theoretical framework using Hu, Zernike and Legendre moment to

155 restore blurred images. The theoretical model is also validated using the image dataset and the results are also

been tested. The result of image dataset is satisfactory for restoring the blurred images. The application is been
 tested on three types of image such as Fingerprint, bird and human face. For majority of the image restoration
 Legendre moments demonstrate good results.



Figure 1: 1 For

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Figure 2:



Figure 3: a

14 CONCLUSION



Figure 4: ?Eq 7



Figure 5: Figure 1 :







Figure 7: Figure 2 . 2 :







Figure 9: Figure 2 . 4 :

	Compare Mome	ents for Blur Ima	ge Restoration	
Q	0	Ω	Q	0
Original Image	Degraded Image Degrade Image for Testing	Hu Restored Image Apply Hu Noment	Zemike Restored Image m & n <u>4 } 0 0011</u> Apply Zemike Moment	Variable Legendre Restored Ima
Angle 4 9	0			Apply Variable Legendre Moment
Length 4 + 3 Add Noise Type - Noise Type melt & pros Noise Density 4 - 2 Mean 4 - 3 Variance 4 - 3	0 900	tatus: Busy		

Figure 10: Figure 2 . 5 : Figure 2 . 6 :

Ι

Capture Agent Type	Range of	
	Blur Value	
	(In MM)	
Gamma Ray Camera	10 to 2	
Ultrasonic Camera	5 to 2.1	
Magnetic Resonance Camera	3.4 to 1	
Computed Thermography Camera	2 to 1.3	
Motion Capture Camera	2.8 to 0.3	
Radio Active Camera	0.5 to 0.1	
V.		

Figure 11: Table I :

 \mathbf{II}

				Year 2016) f (
Input Image	Blur Length	Hu Moment	Zernike	Legendre Moment
		(In %)	$\begin{array}{c} \text{Moment} \\ \text{(In \%)} \end{array}$	(In %)
	10 mm	78	65	53
Fingerprint	20 mm	80	68	58
	30 mm	83	71	68
	10 mm	23	63	71
Bird	20 mm	24	68	75
	30 mm	24	71	79
	10 mm	37	53	81
Human Face	20 mm	41	57	83
	30 mm	53	61	87

[Note: Study of Hu, Zernike and Legendre moment based on K value in Eq. 15. For dataset of fingerprint, bird and human face]

Figure 12: Table II :

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