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# AN APPROACH TO IDENTIFY NOISE IN IMAGES AND APPLY RESTORATION TECHNIQUE TO REMOVE NOISE

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Abstract: Increase in road transportation in India demands present research movements to provide safe road. Maintenance of road and traffic control is one of the important issue for country like India. The safety and traffic situations are depend partly on the road conditions, for Indian road. Hence, the preventive measures for the road conditions are significant factor to consider for road authorities. Various researchers focus on automation of road maintenance. We have studied accuracy will depend on the quality of the input images. Input data can easily be improved by using various high definition capture devices which lead towards higher cost factor and low value propositions. Thus, improving the input image quality is the basic need to improve the accuracy for detecting the road conditions. Here, we studied different types of noise and effective way of noise removal using Legendre restoration technique.

Keywords: noise, restoration, road, maintenance

#### I.INTRODUCTION

The digital images are one of the common sources of data representation and transmission. There is continuous effort carried out to process and present superior images. In image processing facial image is the main research from past 2 decades. Much of the work processed in recognition of facial images for security, capable data management, automations and emotion detection. However the collection of input images strongly depends on capture device. Thus the researchers are making continuous efforts to enhance image quality. The major drawback for recognition of facial images consists of random distortion and blurring due to the noise [1], [4].

## II. LITERATURE SURVEY

Major sources of the distortion, blurriness are environmental and human errors and quality of capturing device. Blurriness in images cause due to focal length of the capture devices, when not adjusted properly with the sync of distance between the object and the length. Captured object in a time irrelevant scale mapped with the capture speed of the agent stay away from the blurriness studied by [2]. Environmental and human causes problem when object is in advanced order of colour range, background of lower order of colour range. Thus to remove the effect of blurriness of the image, momentary calculation algorithms provides efficient results. Moments are the numeric values used to signify the nature of any functions and recognize the significant properties proposed by [2],[3]. Moment correspond to a normally calculated average value denoted as image descriptor used for reconstruction or other recognition tasks. The most widely used moments

algorithms are Hu moment, Zernike moment and Legendre algorithms.

## III. DIFFERENT TYPES OF NOISES

Noise is random signal and can destroy some information present in images. Distortion in images caused by types of noise such as Gaussian noise, Poisson noise, Speckle noise, Salt and Pepper noise.

# A. Salt and Pepper:

Salt-and-pepper noise occur by difficulty in the image signal present with white and black pixels. Noise reduction technique for salt and pepper noise is median filter [5]or amorphological filter[6] worked with salt noise or pepper noise removal using contraharmonic mean filter[7]. Removal of impulse noise using median median filters causes blurring and results in distorted edges and lacks quality. [8] proposed method based on thresholding of pixels for noise removal by median filter using selective pixels that are not noise themselves. Results obtained were compared with usual median filter and center weighted median (CWM) filter and achieved better results.[9]proposed Cellular Automata computational tools for nature and bioinspired computing for efficient approach of grayscale images. Used 2D Moore neighborhood, proved worthy for removal of high-noise levels.[10] modified Adaptive Threshold Median Filter for salt and pepper noise removal on color images of low and high-density noise levels. Results compared with Peak Signal Noise Ratio values and Mean Square Error method and achieved better results.

#### B. Poisson:

Poisson distribution is due to photon arrival process and uncertainty associated with the measurement of light. Magnitude of poisson noise is signal dependent and main source of noise apart from low-light conditions.

#### C. Speckle:

Speckle is a granular noise exist in active radar, synthetic aperture radar, medical ultrasound and optical coherence tomography images. On synthetic or natural surfaces noise is extremely rough on the scale of the wavelength within the resolution cell. Constructive and destructive interference identified as bright and dark dots in the image and distributed as scatterers[11]. Speckle noise in conventional radar results from random fluctuations in the return signal from an object and raise the mean grey level of a local area[12].This noise causes troubles understanding[12][13]. It is caused by coherent processing of backscattered signals from multiple distributed targets. Speckle noises are signals from elementary scatterers, the gravity-capillary ripples, and manifests as a pedestal image, beneath the image of the sea wave[14][15].

#### D. Gaussian:

Data acquisition is one of the source of Gaussian noise. Factors leads to Gaussian noise are noise present in sensor of illumination and its temperature, and connected electronic circuit. [16] Gaussian, is additive, independent at pixel, and at signal intensity.[17]Amplifier noise is caused due to image sensor and constant noise level in dark areas.[18] In color cameras where added amplification used in the blue color channel as compared to green or red channel, leads to noise in the blue channel.[19] proved at higher exposures sensor noise is subject by shot noise, not Gaussian and not independent of signal intensity.[20] acquired data from remote sensing satellites and processed with white Gaussian noise of zero mean and constant variance. Noise denoised through Legendre Fenchel Transformation, and compared with results obtained by Euler-Lagrange ROF model and proved highly convergent and a lesser amount of time.

## IV. ARCHITECTURE OF PROPOSED SYSTEM

We have considered images of Indian road network and degrades image with salt and pepper, Gaussian, Poisson and Speckle Noise. Images are restored with Legendre Moment. We have noticed for digital images Legendre moment outperforms with good results.(Fig 1)

#### V. IMPLEMENTATION

For implementation of work we used images captured from the Indian road of Maharashtra region collected from public work department.

Input images degraded with estimated angle, length with noise such as salt and pepper, Gaussian, Poisson, Speckle and restored images SNR calculated consider ing 10 images. We have observed signal to noise ratio reduced and better results derived as compared to the noisy images SNR.

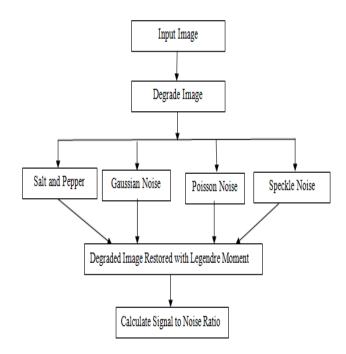


Fig1: Architecture of proposed system







Fig2: Input Images

|          |           |           |         | l, International Jou | rnal of Advano |
|----------|-----------|-----------|---------|----------------------|----------------|
|          |           | able1: Re |         |                      | 1              |
| Sr.      | Noise     | Angle     | Le      | Restoration          | S              |
| No       | Type      |           | ng      | Technique            | N              |
|          |           |           | th      |                      | R              |
| Im1      | Salt and  | 180       | 53      | Legendre             | 0.0012         |
|          | Pepper    |           |         | Moment               |                |
|          | Gaussian  | 178       | 13      | Legendre             | 0.0012         |
|          |           |           |         | Moment               |                |
|          | Poisson   | 180       | 75      | Legendre             | 0.0012         |
|          |           |           |         | Moment               |                |
|          | Speckle   | 45        | 3       | Legendre             | 0.0012         |
|          | Specific  |           |         | Moment               | 0.0012         |
| Im2      | Salt and  | 180       | 42      | Legendre             | 0.0013         |
| 11112    | Pepper    | 100       | 72      | Moment               | 0.0013         |
|          | Gaussian  | 179       | 3       | Legendre             | 0.0013         |
|          | Gaussian  | 1/9       | 3       |                      | 0.0013         |
|          | D :       | 100       | 40      | Moment               | 0.0012         |
|          | Poisson   | 180       | 40      | Legendre             | 0.0013         |
|          |           |           |         | Moment               |                |
|          | Speckle   | 45        | 34      | Legendre             | 0.0013         |
|          |           |           |         | Moment               |                |
| Im3      | Salt and  | 180       | 43      | Legendre             | 0.0013         |
|          | Pepper    |           |         | Moment               |                |
|          | Gaussian  | 179       | 3       | Legendre             | 0.0013         |
|          |           |           |         | Moment               |                |
|          | Poisson   | 90        | 17      | Legendre             | 0.0013         |
|          | 1 0133011 |           | 1,      | Moment               | 0.0013         |
|          | Speckle   | 179       | 3       | Legendre             | 0.0013         |
|          | Speckie   | 1/9       | 3       |                      | 0.0013         |
| T 4      | 0.1, 1    | 100       | 27      | Moment               | 0.0041         |
| Im4      | Salt and  | 180       | 37      | Legendre             | 0.0041         |
|          | Pepper    |           |         | Moment               |                |
|          | Gaussian  | 154       | 3       | Legendre             | 0.0041         |
|          |           |           |         | Moment               |                |
|          | Poisson   | 180       | 19      | Legendre             | 0.0041         |
|          |           |           |         | Moment               |                |
|          | Speckle   | 180       | 19      | Legendre             | 0.0041         |
|          |           |           |         | Moment               |                |
| Im5      | Salt and  | 180       | 34      | Legendre             | 0.0019         |
|          | Pepper    |           |         | Moment               |                |
|          | Gaussian  | 116       | 28      | Legendre             | 0.0019         |
|          | Gaassan   | 110       | 20      | Moment               | 0.0019         |
|          | Poisson   | 190       | 21      | Legendre             | 0.0019         |
|          | 1 0188011 | 180       | 21      |                      | 0.0019         |
|          | Cm c :1-1 | 125       | 2       | Moment               | 0.0010         |
|          | Speckle   | 135       | 3       | Legendre             | 0.0019         |
| <b>.</b> | 9.1       | 460       |         | Moment               | 0.00::         |
| Im6      | Salt and  | 180       | 57      | Legendre             | 0.0014         |
|          | Pepper    |           |         | Moment               |                |
|          | Gaussian  | 178       | 17      | Legendre             | 0.0014         |
|          |           | <u> </u>  |         | Moment               |                |
|          | Poisson   | 180       | 80      | Legendre             | 0.0014         |
|          |           |           |         | Moment               |                |
|          | Speckle   | 178       | 3       | Legendre             | 0.0014         |
|          | ~         |           | -       | Moment               |                |
| Im7      | Salt and  | 90        | 11      | Legendre             | 0.0018         |
| 1111/    | Pepper    | 70        | **      | Moment               | 0.0010         |
|          |           | 32        | 27      |                      | 0.0018         |
|          | Gaussian  | 32        | 27      | Legendre             | 0.0018         |
|          | D :       | 100       | 20      | Moment               | 0.0010         |
|          | Poisson   | 180       | 20      | Legendre             | 0.0018         |
|          |           |           |         | Moment               |                |
|          | Speckle   | 65        | 3       | Legendre             | 0.0018         |
|          |           |           | <u></u> | Moment               |                |
| Im8      | Salt and  | 180       | 21      | Legendre             | 0.0023         |
| I        | D         | I         | ĺ       | Manage               | 1              |

|      | Gaussian | 179 | 3  | Legendre | 0.0023 |
|------|----------|-----|----|----------|--------|
|      |          |     |    | Moment   |        |
|      | Poisson  | 180 | 19 | Legendre | 0.0023 |
|      |          |     |    | Moment   |        |
|      | Speckle  | 153 | 3  | Legendre | 0.0023 |
|      |          |     |    | Moment   |        |
| Im9  | Salt and | 180 | 21 | Legendre | 0.0011 |
|      | Pepper   |     |    | Moment   |        |
|      | Gaussian | 179 | 38 | Legendre | 0.0011 |
|      |          |     |    | Moment   |        |
|      | Poisson  | 1   | 28 | Legendre | 0.0011 |
|      |          |     |    | Moment   |        |
|      | Speckle  | 179 | 3  | Legendre | 0.0011 |
|      |          |     |    | Moment   |        |
| Im10 | Salt and | 180 | 38 | Legendre | 0.0016 |
|      | Pepper   |     |    | Moment   |        |
|      | Gaussian | 180 | 51 | Legendre | 0.0016 |
|      |          |     |    | Moment   |        |
|      | Poisson  | 180 | 25 | Legendre | 0.0016 |
|      |          |     |    | Moment   |        |
|      | Speckle  | 180 | 37 | Legendre | 0.0016 |
|      | _        |     |    | Moment   |        |

#### VI. CONCLUSION AND FUTURE WORK

In the study of road network safety, maintenance of the roads so far follows manual approach and needs to be improved to fast up process with automation. In the study we noticed one of the significant improvements on automatic detection of the road condition depends on the image quality captured. So for further analysis input image should be free of noise and blur factors. Hence, this work demonstrates an automatic approach to detect and restore noise within images. The result obtained from this work is noteworthy and provides better analysis of road conditions and processing.

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