

# TO ENHANCE THE PNLM FILTERING SYSTEM IMAGES FROM MRI TO DENOISE

**Vishal Agnihotri**  
Research Scholar

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## Abstract

*The procedure that can cycle data stored as pixels is picture management. Clinical pictures with raised noise are what the MRI scans look like. In the past, many channels for denoising photos were created. This research examines the different image denoising and separation methods. Testing has shown that the fix-based approach yields the best picture denoising outcomes in terms of PSNR and MSE. In MRI pictures, low distinction and noise are frequent issues, especially when imaging the heart and brain. Because a skilled radiologist needs to get a precise judgement, its utilisation is necessary in a large clinical organisation. The partition of characteristics, picture ordering, restoration of three-dimensional pictures, and enrollment are all greatly hampered by this noise. Noise in MR pictures will change the motivation for each pixel to be adequate and its stage. As a result, the visual quality declines and testing is necessary to precisely diagnose the sickness. Top-notch photos of human tissues and organs require advanced clinical picture handling.*

**Keywords:** *Denoise, MRI Images, PNLM, PSNR, Rican Noise*

## I. INTRODUCTION

Image handling is the practise of employing various approaches to enhance or separate key information from an image. When an image is given as input, the result can either be a more effective de-noising technique or a few fragments of that image. In this field, both human and automatic picture handling processes are applied. Advanced picture handling is the term for computer-based photo editing. Pre-handling, expansion, and data extraction are the three

fundamental cycles that apply to an information being handled in this discipline. Scanners, cameras, and video cards are the three components of technology that are most essential for gathering images from multiple sources. Image handling carries out extra numerical duties to the photos while also enhancing the difference, detecting edges, calculating force, and guaranteeing a low cost speculation. [1] The issues of picture processing go beyond improving the brightness of pictures or changing the spatial intent of ordinary camera images. Picture handling is a function that may be found in many programmes. A few notable applications of photo processing include tracking moving objects, remote detection, intelligent transportation systems, biological imaging, and automated visual inspection. MRI, which stands for attractive reverberation imaging, is a method for turning raw data into images before seeing them in a repeating spatial region.

The technique and interaction known as "clinical imaging" is used to monitor and communicate visual representations of a body's interior organs and tissues for the purposes of clinical investigation and clinical therapy. Also, it gives a visual picture of how certain internal organs or tissues work. In order to improve general wellbeing for all demographic groups, it therefore plays a significant role. Since it may sometimes show interior structures that the skin and bones can't see, clinical imaging is very useful in diagnosing and treating sickness. It is a type of natural imaging that builds a knowledge base of fundamental physiology and living systems to look for anomalies. This technique combines many imaging advancements, including X-ray radiography, appealing reverberation imaging, clinical ultrasound, endoscopy, elastography, thermography, material imaging, clinical photography, and atomic medicine practical imaging techniques like PET and SPECT. Because to the various repeated patterns visible in both natural and medical images, the NLM filter, first proposed by, has attracted a lot of interest in demising, particularly MR images. Conventional MRI demising algorithms, which were first developed to eliminate Gaussian noise from a picture, were unable to recognise rician noise. Later, new techniques like non-local means (NLM), wavelets, partial differential equations (PDE), and maximum likelihood (ML) were suggested. These techniques were improved even further to account for Rician noise. This paper discusses the removal of rician noise (corrupted data) in MR pictures using the NLM filter in addition to a description of the different picture modalities, noise kinds, and filtering strategies.

For the recovery of the fine features, picture data, and extraction of hidden information in applications of medical imaging, noise reduction from an image is crucial. The process of medical diagnosis is impacted by these noisy, distorted MR images. In general, removing noise from a picture is an important topic of study and is accomplished in a variety of ways. Many previously suggested tools are used on the abandoned MR pictures. In order to balance the denoising, it is crucial to maintain the edges in MR pictures since they play a vital role in the images. The noise-suppression method must, however, avoid damaging or degrading the important aspects of a picture. Buades suggests using a filter called Non Local Means (NLM). Several different denoising techniques have been used to reduce noise from photographs; each of these techniques only considers surrounding local pixels. While larger structures are kept, smaller ones are seen as noise and deleted from a picture. The noise must be eliminated using the NLM filter since photographs contain redundant information.

## II. REVIEW OF LITREATURE

The Picture Denoising test is a crucial test that researchers take into consideration, according to JannathFirthouse.P, et al. (2016). In this work, MRI images were processed using a contour let space for information picture decay to remove Gaussian and Dot noises. [4] They applied the previously specified safety method to safeguard the curves and edges. The noise impacts might be further reduced in this study by using limit approaches like Bayes, Neigh, and Block Shirnk throughout the degradation cycle. Many parametric criteria are developed to ensure that the improvement promised by the recommended technique has been realised.

According to FarhaFatina Wahid et al. (2017), the most widely used innovation in the field of clinical imaging technology is attractive reverberation imaging (MRI). The combination of at least two denoising calculations is now the key component. The major goal of this work is to integrate two often employed denoising techniques, or the idea of local and non-local filtering in MRI images[5]. In this investigation, the fundamental Non Local Means filter (NLM) as well as the local fluffy filter were both employed. Also, they combined two techniques in the modified picture using the Non Subsampled Contourlet Transform (NSCT). The additional material noise that had damaged a handful of MRI pictures has now been completely removed. Tests to evaluate how well the suggested approach was explained were completed. The

obtained results demonstrated that, in comparison to other options, the suggested approach was practicable.

A.G. The age of clinical analysis saw the introduction of Magnetic Resonance Imaging (MRI), a recently invented noninvasive technique that has since been warmly accepted. Here is yet another example of an encounter that is judged puzzling for the sake of a picture analysis. Examples are division, procreation, and enlisting. Noise reduction is therefore considered the preprocessing stage in MR pictures before moving on to picture assessment. [6] Two-layered (2D) photography has provided a useful understanding for difficult commodities. This calls for the depiction of numerous real objects in 3D computer. In this study, they improved the denoising and division procedures in light of fractal and morphological techniques.

R. The analysis of clinical images by the discovery system is primarily reliant on the picture denoising, according to Sujitha et al 2017 's presentation. How to eliminate noise from the original sign is the major problem that the experts are focusing on. One of the many tactics that have been put out to date, denoising procedures, each has advantages and disadvantages. The suggested technique was applied during the clinical conclusion. Also, the inspection area of these images was denoised, which improved the image's quality and clinical limits. A wavelet-based thresholding technique for noise concealment and denoising of MRI image images was suggested in this research. A few parametric attributes are provided for the analysis of the presentation level of the recommended technique.

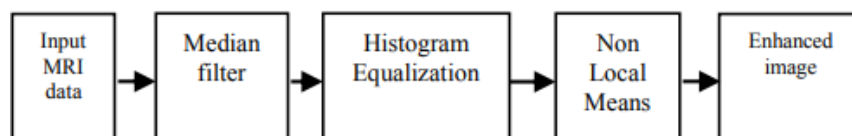
Magnetic Resonance Imaging (MRI), a recent development in therapeutic imaging that permits the collecting of very exact images from the human mind, was introduced by Hanafy M. Ali et al. in 2017. The MRI technology is used to study the numerous human organs. The images that were successfully analysed recall a few different types of sounds that affect how those sights are interpreted. In this work, they changed a computation for an intermediate filter. The numerous noises mentioned above in this study are used to improve the MRI image. They removed unnecessary sounds from the MRI images using the filters specified in this paper. They added the noise thickness to the MRI image to account for the display and evaluate the filters.

Dongsheng Jianga et al. (2018) By concluding the previously stated method of denoising magnetic resonance (MR) images, it should be possible to work on the character of the

produced picture. They suggested utilising a convolutional neural network to reduce MRI Rician noise. A brain network with 10 convolutional layers and multi-channel, persistent learning were also suggested. In order to assess the suggested strategy, tests were done on both fake and real 3D MR data. The findings obtained demonstrate that, in comparison to other methods, the suggested strategy is suitable in terms of the biggest sign to noise ratio and the global design closeness list. In the unique situation where there is no noise level barrier while utilising approaches in two datasets, the provided method will perform well. As a result, the recommended strategy showed strong generalizability.

### III. METHODOLOGY

The recommended boosting process is broken down into three steps. The first stage in any noise reduction picture processing approach is pre-processing. The median filter is used in the first stage of pre-processing because it provides better pre-processing with less edge blurring. The repeatability, lack of sensitivity to specific noise spikes, capacity to properly attenuate impulsive noise, and absence of visibly blurred edges of median filtering are its benefits.



**Figure 1:** Proposed approach

The second stage, referred to as the difference upgrade, concentrates on the appearance of minute highlights and aids in the differentiation of the CSF, white matter, and black matter. This is accomplished through histogram adjustment since it is effective and transparent. The last stage of denoising employs the non-iterative Non Local Means filter, which jells fine designs and reduces identical sounds.

By exposing the subject to both homogeneous and non-homogeneous Gaussian and Rician noise's properties, this hybrid technique improves visual differentiation[7]. The technique also gets rid of the propensity effect caused by uncorrelated Rician noise.

The moving window rule, which compares the entire picture with the section, implements the middle filtering by replacing the computed middle for the middle pixel value of each window.

[8] The created histogram is now extended to a new, unique reach by the histogram balancing, which further contributes to the total difference. According to the Non Local Means filter, each tiny window (fix) in a distinctive image has numerous indistinguishable windows in a related image, which relies on the likelihood that the image includes a critical quantity of self resemblance. In order to calculate the pixel loads for noisy picture filtering, non-local means looks at entire repairs. Using a highlights space composed of picture slopes and the local mean value, it selects the nonlocal loads according to distance (Buades et al, 2005). The procedure increases accuracy and decreases discrepancies between pixels in the picture by using data that is close to comparability. The non-local implies denoised image's pixel I is classified as

$$nlm(I_n)(i) = \sum_{j \in I_n} w(i, j) I_n(j)$$

where  $I_n(i)$  represents the pixel being filtered,  $I_n(j)$  represents each pixel in the noisy image, and weights meets the requirement and. Hence, the weight depends on how similar the neighbourhood windows  $N(i)$  and  $N(j)$  of the two pixels are to one another. A diminishing function of the weighted Euclidean distance is used to measure how similar it is. The weight is now described as,

$$w(i, j) = \frac{1}{Z_i} e^{-\frac{\|I_n(N_i) - I_n(N_j)\|^2}{h^2}}$$

where  $Z_i$  assures that by serving as the normalisation constant  $\sum$ .

$$Z_i = \sum_{j \in I_n} e^{-\frac{\|I_n(N_i) - I_n(N_j)\|^2}{h^2}}$$

The rate at which an extraordinary capacity degrades is governed by the smoothing boundary  $h$ . The equation is  $h = (k)^2$ , where  $k$  is the noise standard deviation and filtering level. The inclination impact is now having an impact on the Rician noise. [9] To reduce Rician noise, a clear inclination treatment is provided to the NLM filter. By subtracting the noise preference from the squared value of NLM, the unprejudiced NLM is produced. It is given by

$$UNLM(I_n) = \sqrt{\max(NLM)(I_n)^2 - 2\sigma^2}, 0$$

In order to prevent undesirable characteristics as a result of the deduction activity, the restored worth of the amount is zero and serves as the upper limit of the worth after the deduction.

Lastly, the strategy's presentation is approved using a number of measures, including RMSE, PSNR, and SSIM. The NLM filter can be quite effective since Rician noise in MRI pictures is a persistent issue. This is accomplished by weighting nonlocal pixels according to how similar they are to the goal pixel and averaging nonlocal pixels. The size of the hunt window, the area or way window, and the amount of filtering must all be changed in order to employ the NLM filter. [10] Basic factors that affect how effectively the computation works include the size of the inquiry and the closeness or fix windows. Using covering sets of pixels, the block-wise NLM filter is carried out. The benefit of NLM is that the procedures do not use iterative cycles. In terms of execution, the NLM filter outperforms earlier methods like absolute variety, wavelet thresholding, and anisotropic dispersion filtering. As it examines both the mathematical structure of the entire area as well as the dark level at a single location, the NL means filter provides a more thorough analysis than other area filters.

#### **IV. RESULTS AND DISCUSSIONS**

##### ***a) Datasets***

The approach has been evaluated against current denoising approaches using both real and reenacted MR datasets. The McConnell Cerebrum Imaging Center of the Montreal Neurological Organization, McGill College, collects the reconstructed mind MRI datasets. The informative index is made up of 181 x 217 x 181 voxel T1weighted Pivotal, T2weighted Hub, and PD images. The ground truth picture from the brainweb database was subjected to various levels of Rician and Gaussian noise, ranging from 3% to 12%, to confirm the results[11]. Gaussian noise was applied to the real and imaginary components before the image's size was selected. The images from the clinical informative indexes were obtained from <http://www.osirix-viewer.com/datasets> and were taken with a Philips Clinical Frameworks 1.5T Scanner. Here, spin echo (SE) sequences with long reiteration time (TR) and short echo duration are used to acquire the images (TE). T1 weighted Hub MR images of a typical cerebrum with TR = 449 ms, TE = 10 ms, 5 mm thickness, and 512 512 target are used to validate the results.

##### ***b) Validation Strategies***

The retention of edge highlights and the presence of curios are two factors that are occasionally used to evaluate a picture's quality. It regularly undergoes external evaluation. The algorithmic noise, often referred to as strategy noise or surviving picture, is shown by the difference between the initial picture and its denoised picture. It is important to verify that any physical data traces that were eliminated during denoising should sound as much like background noise as is humanly feasible [12]. The denoising display improves when there are less apparent image structures. Here, many measures are used to assess the feasibility of the procedure:

- 1 Mean Square Error (MSE)
- 2 Root Mean Square Error are two examples (RMSE)
- 3 Maximum Signal-to-Noise Ratio (PSNR)
- 4 Index of structural similarity (SSIM)

The MSE quantifies the intensity of the error signal through the use of a formula.

$$MSE = \frac{1}{mn} \sum_{m=1}^m \sum_{n=1}^n [I(i,j) - I_d(i,j)]^2$$

The symbols  $I(i, j)$  and  $I_d(i, j)$ , where  $mn$  is the image dimension, respectively, depict the intensities of the pixels  $I_j$  in the original picture and the demonised image. The RMSE, a gauge of how successfully noise is removed, is calculated using the square root of the MSE. The MSE and RMSE values need to be as little as feasible for better denoising. The formula outputs a decibel value for the Peak Signal to Noise Ratio (PSNR).

$$PSNR = 10 \log_{10} \left( \frac{MAX}{MSE} \right)$$

### c) *Simulation Results and Observations*

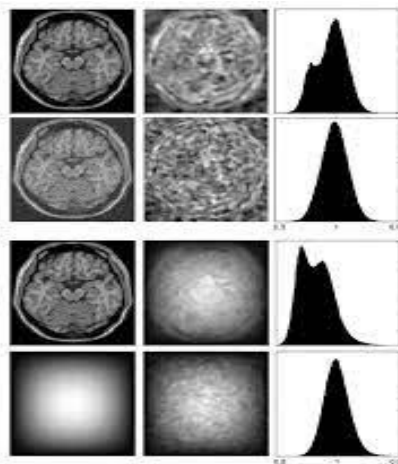
With a 2.40GHz Intel(R) Centre i5 computer processor and 4.00GB of RAM, the discoveries are verified. MATLAB R2012a running on Microsoft Windows 7 64-bit is the programming used to validate the philosophy. The demising results for the T1 weighted, T2 weighted, and PD hub images that were distorted by spatially homogeneous and no homogeneous Gaussian and Random noise at noise levels ranging from 3 to 12% are shown in Figs. 2 to 5. The results



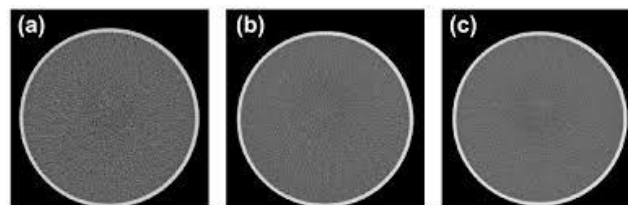
of the exploratory research demonstrated that the method outperformed alternative tactics in both objective and quantitative evaluations.

The 9% homogeneous Gaussian noise in Fig. 2 has distorted the T1-w contrast enhanced image. ADF is still present in the baseline image on the left and NLM is still present in the image with unique differentiation improvement in the top column on the right. Left centre column: ADF approach; right column: suggested NLM technique.

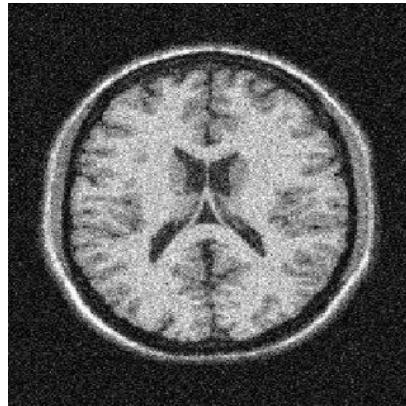
The 7% non-homogeneous Gaussian noise distorts the T2-w contrast-improved image in Fig. 3. The top line on the left shows the initial differentiation-improved image. Gaussian noise that is 7% non-homogeneous degraded the image to the right. The centre line displays the ADF approach. The base line displays the suggested NLM method..



**Figure 2:** Validation with homogeneous Gaussian noise

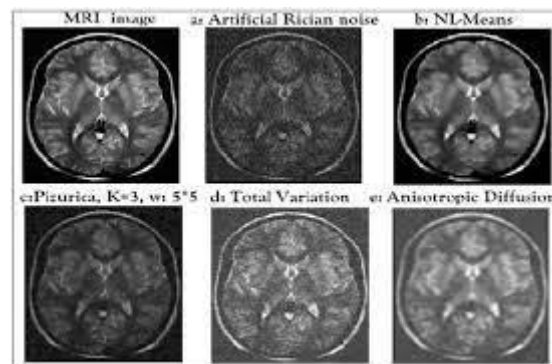


**Figure 3:** Validation with non homogeneous Gaussian noise



**Figure 4:** Validation with homogeneous Rician noise

Figure 3 shows a PD contrast-upgraded picture that has 9% consistent Rician noise contamination. 9% homogenous Rician noise has been used to alter the picture on the top line to the right. Left baseline: ADF is still in place; centre: NLM method; right baseline: UNLM approach. The NLM approach is in the left column, while the ADF approach is in the right column.



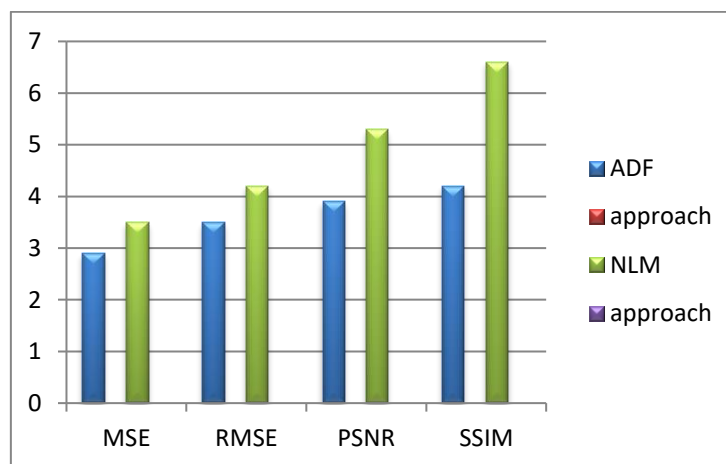
**Figure 5:** Validation with non homogeneous Rician noise

There is 11% non-homogeneous Rician noise contamination in the T2-w contrast-upgraded picture in Fig. 4. ADF still in use; NLM method still in use; UNLM technique still in use; base column, left; right. Updated picture with a noticeable alteration in the top leftmost column. 12% Rician nonhomogeneous noise destroyed. ADF strategy is in the left-center column. Attention: NLM technique. Right: The suggestion from UNLM. To complement the methodology's demonstration, three separate clinical real-world T1w pictures with a 5mm cut thickness are also employed.

The original picture in the example below has been warped by 5% homogenous Gaussian noise. The recommended method outperformed the ADF at high noise levels on both fictitious and actual data, according to the visual correlation of these findings, creating a more comprehensive denoised image in which all of the distinguishing traits and subtle essential details are strongly retained. Table 1-3 lists the presentation metrics for the T1 weighted, T2 weighted, and PD critical pictures with different noise levels. Table 3 shows usual information.

Method	MSE	RMSE	PSNR	SSIM
ADF approach	2.9	3.5	3.9	4.2
NLM approach	3.5	4.2	5.3	6.6

**Table 1 :** Comparison based on performance metrics for T1w image corrupted by 9% homogeneous Gaussia noise.

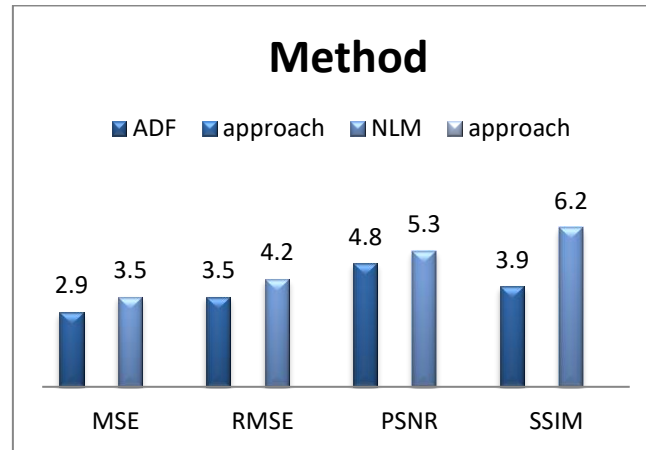


**Figure 6:** Comparison based on performance metrics for T1w image corrupted by 9% homogeneous Gaussia noise

Method	MSE	RMSE	PSNR	SSIM
ADF approach	2.9	3.5	4.8	3.9
NLM	3.5	4.2	5.3	6.2

approach				
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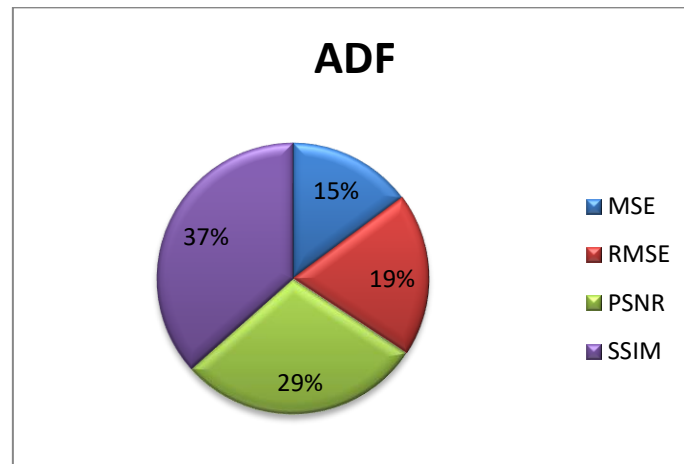
**Table 2:** Comparison based on measurements for a T2w image that has 7% non-homogeneous Gaussian noise in it



**Figure 7:** Based on performance measurements for a T2w image with 7% non-homogeneous Gaussian noise, a comparison is made.

Method	MSE	RMSE	PSNR	SSIM
ADF	3.2	4.2	6.3	7.9
approach				
NLM	3.9	5.6	7.2	8.2
approach				

**Table 3:** Comparison based on performance metrics for PD image corrupted by 9% homogeneous Rician noise



**Figure 8:** Comparison based on performance metrics for PD image corrupted by 9% homogeneous Rician noise

When the amount of noise grows, the proposed NLM filter performs noticeably better than the following denoising techniques. The suggested filter performs better than elective denoising techniques, as evidenced by higher PSNR and SSIM values..

## V. CONCLUSION

The filtering techniques are examined in the study that we have suggested. The most recent method is suggested for denoising MRI images with raisin-like noise. It uses PNLM, an improved version of NLM. In order to contribute to the suggested NLM filtering strategy, pictures are taken and utilised. This allows for an approximation of the noise levels in the captured MRI pictures. [13] A histogram levelling and NLM-based MRI improvement technique has been put out in this review. The photos have been subjected to varying spatially homogeneous and nonhomogeneous Gaussian and Rician noise conveyances. ADF based on MSE, RMSE, PSNR, and SSIM is contrasted with the suggested method's presentation. [14] While addressing Rician noise, the NLM filter with predilection adjustment (UNLM) performs better than the NLM filter without inclination.

## VI. FUTURE SCOPE

The future course of events will depend on the expansion of this technique as a preliminary step for additional image handling methods, such as segmenting and enrolling patients with cerebrum cancer. For photos with both Gaussian and Rician noise contamination, the

continuing NLM execution works incredibly well, but the computing overhead is significant because every pixel needs to be denoised utilising a number of essential operations. [15] The equation  $O(QRM)$ , which stands for the number of pixels in the picture, the number of pixels in the search window, and the number of pixels in the likeness window, provides its representation. By reducing the dimensions, it is feasible to lower the processing cost of the technique.

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