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Empirical Wavelet Transform Method for Enhancement of Medical Image Fusion

*¹Nelaturi Nagendra Reddy, ²Bitra Jayalakshmi, ³Goli Srinivasa Rao

¹Bapatla Engineering College, Andhra Pradesh, India.

²VIT-AP University, Andhra Pradesh, India.

³Vignan University, Andhra Pradesh, India.

*Corresponding author Email: jayanagendra43@gmail.com

Abstract. *The process of creating an image's emulsion is selecting the crucial details from numerous images and combining them into smaller images, often one bone. In the areas of satellite imaging, remote seeing, target shadowing, medical imaging, and many other areas, image emulsion is quite useful. This design tries to illustrate how Empirical Wavelet transfigures work when used with the Simple Average Emulsion Rule to emulsify multi-focus images. The suggested approach has been tested using common datasets for merging images with various focal points. Empirical Wavelet Transform is primarily a method that uses an adaptive approach to produce a Multi-Resolution Analysis of the signal. The effectiveness of the suggested approach is calculated in a variety of ways. Visual perception and the evaluation of common quality metrics, such as Root Mean Squared Error, Entropy, and Peak Signal to Noise ratio, are used to compare the performance of the proposed system. The proposed fashion based on the Empirical Wavelet Transform (EWT) outperforms the existing methods, according to the study of the experimental results. According to the suggested criteria, the fused image's entropy should be higher than the component images' because the emulsion's efficiency decreases as entropy increases. This technique takes MRI and CT scans into account.*

1. INTRODUCTION

Image fusion is a method for combining the features from numerous images taken by various sensors into a single image. All pertinent information from the input images is included in the fused output image. Multiple methods are known in the literature to accomplish the objective of multi-focus picture fusion, which is to integrate various images with varied focal points into a single image. Navigation guiding, satellite imaging for remote sensing, object detection and recognition, military and civilian monitoring, etc. all heavily rely on image fusion. Image fusion algorithms can be classified into different levels: pixel, feature, and decision levels [1-5]. It is possible to separate image fusion techniques into two categories, such as spatial domain and transform domain. Discrete Wavelet, Fast Intensity Hue Saturation (FIHS) Image fusion is a method for combining the features from numerous photos taken by various sensors into a single image. All pertinent information from the input image is included in the fused output image. Multiple methods are known in the literature to accomplish the objective of multi-focus picture fusion, which is to integrate various images with varied focal points into a single image. Image emulsion has recently been employed in a variety of tasks including remote viewing, surveillance, obtaining medical opinions, and photography operations. A multi-focus image emulsion and a multi-exposure image emulsion are two of the most important image emulsion activities in photography. Image emulsion's central concept is the collection of critical and necessary information from the input images into a single image that flawlessly contains all of the information of the input images. The exploration of picture emulsion has been documented in more than 30 research works. In general, there are two components to image emulsion: subjective evaluation standards and image emulsion styles. Cameras that record images and videotape sequences are known as detectors in visual detector networks (VSN). A camera cannot provide a perfect illustration of the scene with all of its details in many VSN activities [5-7]. This is brought on by the cameras' optic lenses' shallow depth of focus. As a result, only the object in the camera's field of view is sharp and focused, and the rest of the image is blurry. VSN uses a number of cameras to take pictures with various depths of field. Reusing the original input photos is crucial to reduce the amount of communicated data due to the enormous quantum of data created by cameras compared to other detectors like pressure and temperature detectors and some restrictions of bandwidth, energy consumption, and processing time. Recent research on multi-focus picture emulsions can be divided into two categories: transfigure and spatial fields. Separate Cosine Transfigure (DCT) and Multi-Scale Transform are often utilised transforms for image emulsion (MST). Deep literacy (DL) has recently flourished in a number of image processing and computer vision operations [7-17].

2. RESEARCH APPROACH

The goal of our research is to help doctors quickly and accurately diagnose disorders by efficiently scanning the patient's internal organs. MRI and CT scan images were taken into consideration for this process as medical imaging. The information provided by these two images varies. According to actual medical opinion, it is both time and labor-intensive and cost-effective to complete the task. The frequency coefficients are used in a prior research of picture fusion called the Discrete Wavelet Transform, but as with any discrete process, noise and distortions that arise throughout the process must be dealt with. We are using the Image Fusion by Empirical Wavelet Transform method to solve these problems. For the process that returns the average of the resolution coefficients, we employed the average fusion rule.

Algorithm

A. Algorithm to read CT Image:

- Step 1:** Start the MATLAB software and open the editor to write the code for image fusion.
- Step 2:** Add the path to read the sub-tools in MATLAB. Create a path to images from the location stored in the device so the reading of the image can be easily done by the path in the program.
- Step 3:** Read the CT image from the stored location in the device using the path provided before.
- Step 4:** Check the image for RGB components and if it possesses them then convert the image to gray level and store the image.
- Step 5:** Convert the Greyscale image into an RGB image and compress an image to 256 bits and store the image.

B. Algorithm to read MRI Image

- Step 1:** Create the path to read the MRI image from the location stored in the device so that the reading of the image can be done easily.
- Step 2:** Read the MRI image from a stored location in the device using the path provided before.
- Step 3:** Check the image for RGB components and if it possesses them then convert the image to gray level and store the image.
- Step 4:** Convert the Greyscale image into an RGB image and compress the image to 256 bits and store the image.

C. Algorithm to perform Empirical Wavelet Transform

- Step 1:** Create the subplots to store the CT and MRI image.
- Step 2:** Apply the wavelet filter to return the lowpass and high-pass Components present in the image for the application of the Empirical Wavelet Transform.
- Step 3:** Apply the EWT to the obtained co-efficient and apply the average fusion rule to them.
- Step 4:** Apply Inverse Empirical Wavelet Transform to resultant co-efficient obtained present in the image.
- Step 5:** Subplot the resultant image and name it a fused image. Apply the entropy metric to the MRI and Fused images.
- Step 6:** Calculate the PSNR and MSE for the Fused Image and write the resultant output images and values.
- Step 7:** End.

In this study, we use MATLAB Tools to read the medical images and apply the Empirical Wavelet Transform to both of them. The formats for both photos should match, such as JPEG, PNG, etc. Because of the Multi-Resolution Analysis found in EWT, as indicated in the diagram, the coefficients contained in the image, specifically the resolution coefficients, will be separated into distinct modes upon application of EWT. Now, we apply the fusion method known as Average to these coefficients of various modes. This computes the average of the coefficients, and after that, we apply the IEWT to this image to produce the final image.

Methodology:

In this paper, we propose a new process of image fusion with the help of a method known as Empirical Wavelet Transform. In this process, we read MRI and CT scans, among other medical images. We take into account these images because CT images allow us to see the bony sections of our body, or how the organs are outlined, but MRI images only show the non-bony elements of our body, such as tissues and cells. This allows us to detect tumours, swelling, and other abnormalities in the body. Because it has a feature called Multi-Resolution Analysis that aids in the division of coefficients on the basis of resolution, the EWT has the main benefit over the existing approach in that it uses the resolution coefficients. The read images should be examined for RGB components, and if any are found, the image should be transformed from RGB to Gray level for effective fusion. This is because RGB components can cause image loss because the colour will have a different spectrum and complicate the analysis of the image. Following conversion, we subject these photos to the EWT and the Average rule. These photos are subjected to Inverse EWT, and the fused image is the outcome of this inversion. The final image from the image fusion using the empirical wavelet transform is taken into consideration.

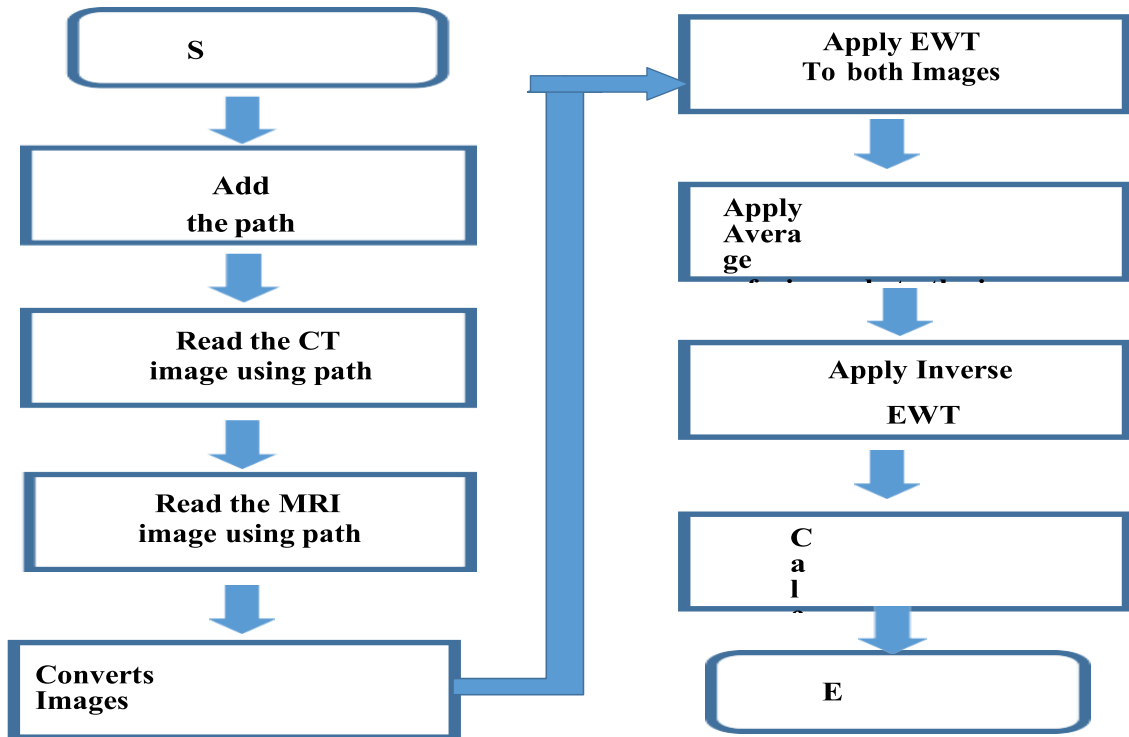


FIGURE 1. Flow chart of EWT.

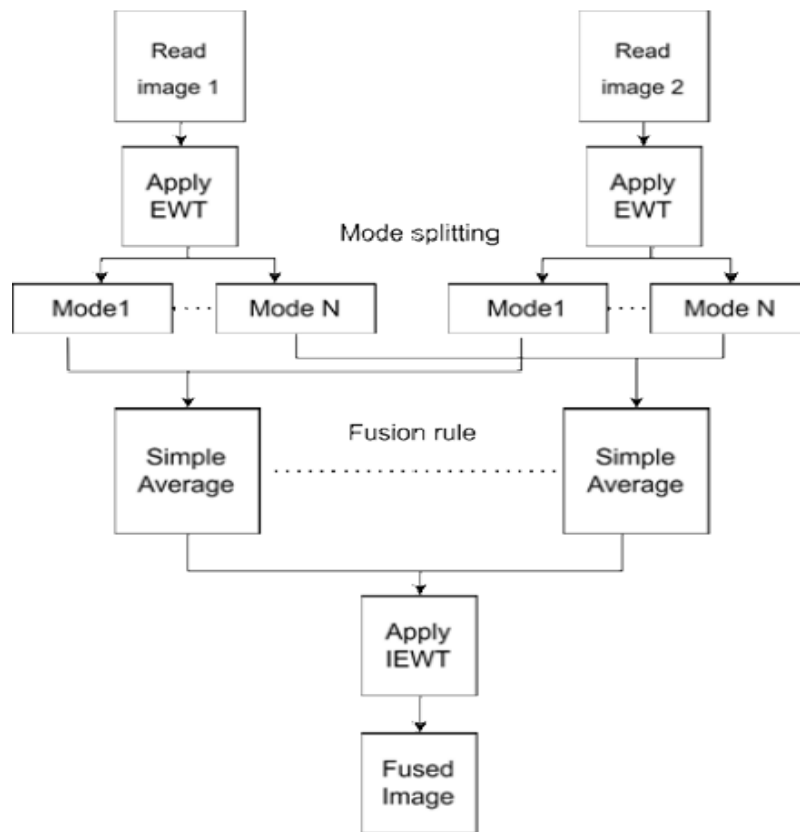


FIGURE 2. Block diagram of EWT.

3. RESULTS

The first step involved in any image fusion process is to read the images since we are using the MATLAB tool, we need to create a path for these images. The image we are reading first is a CT image as shown in the figure

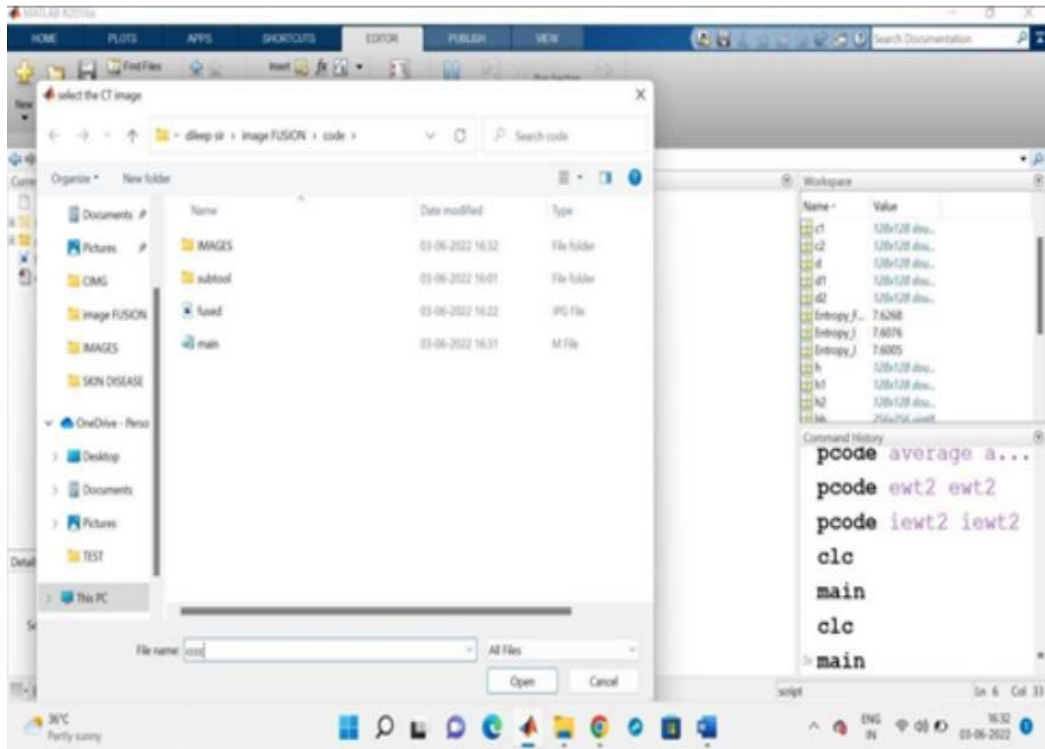


FIGURE 3. Selecting CT Image

The second image we considered is an MRI image same as the previous one we read the image from the stored location using the path provided. The size of the image should not be high. After reading both of these images, we applied the EWT to them, but first we should confirm the format and kind of the image. We cannot execute the fusion if the photos are of different sections of the same organ. The final image, or merged image, is obtained as shown below.

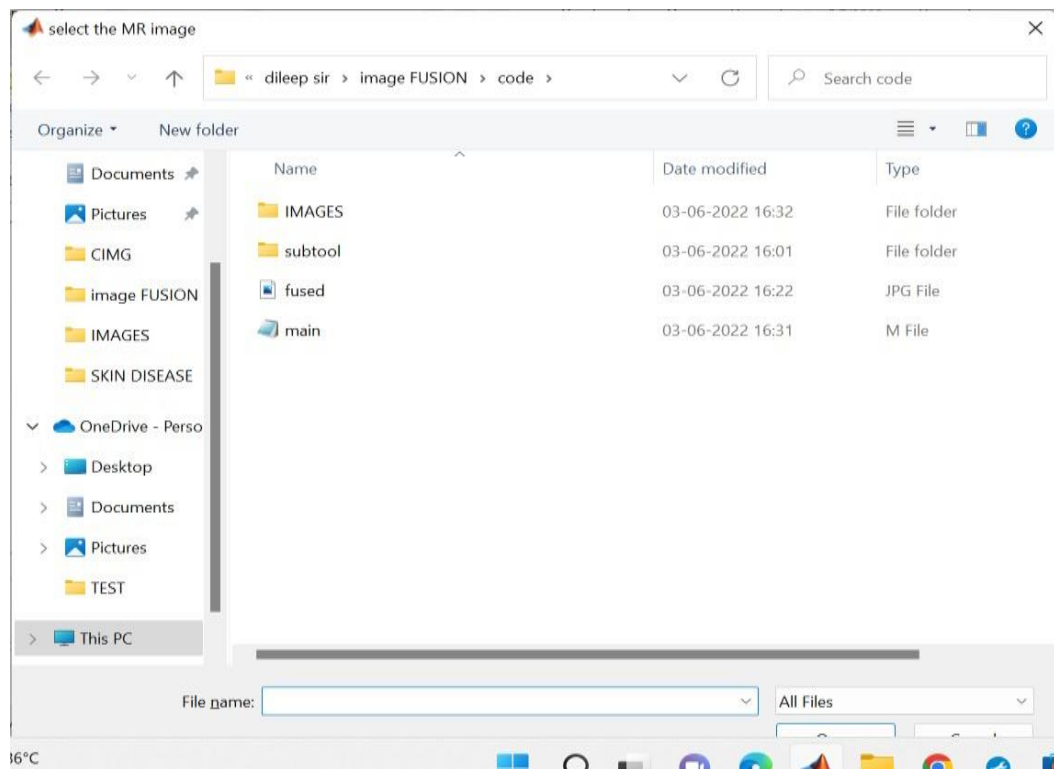


FIGURE 4. Selecting MRI Image

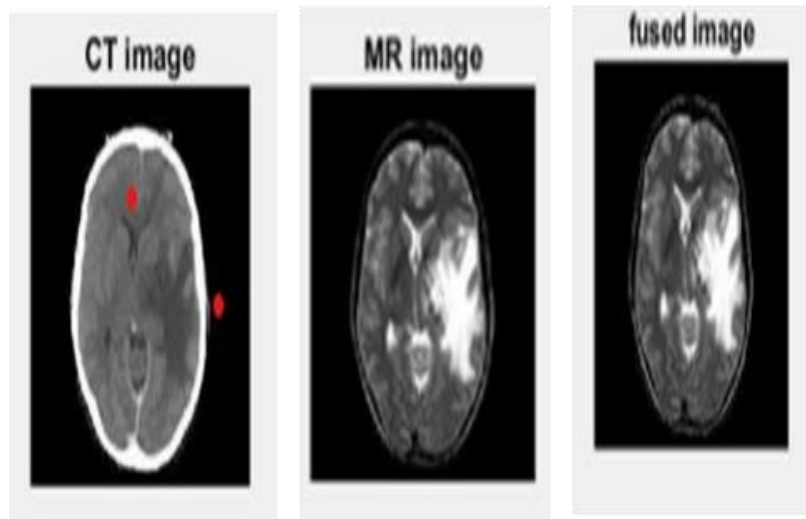


FIGURE 5. Multimodal Image Fusion

The metrics that are calculated are the Entropy for the MRI image, CT image and Fused image in addition we also found PSNR, MSE values for the fused image. For any image the MSE should be below and the Entropy should be high.

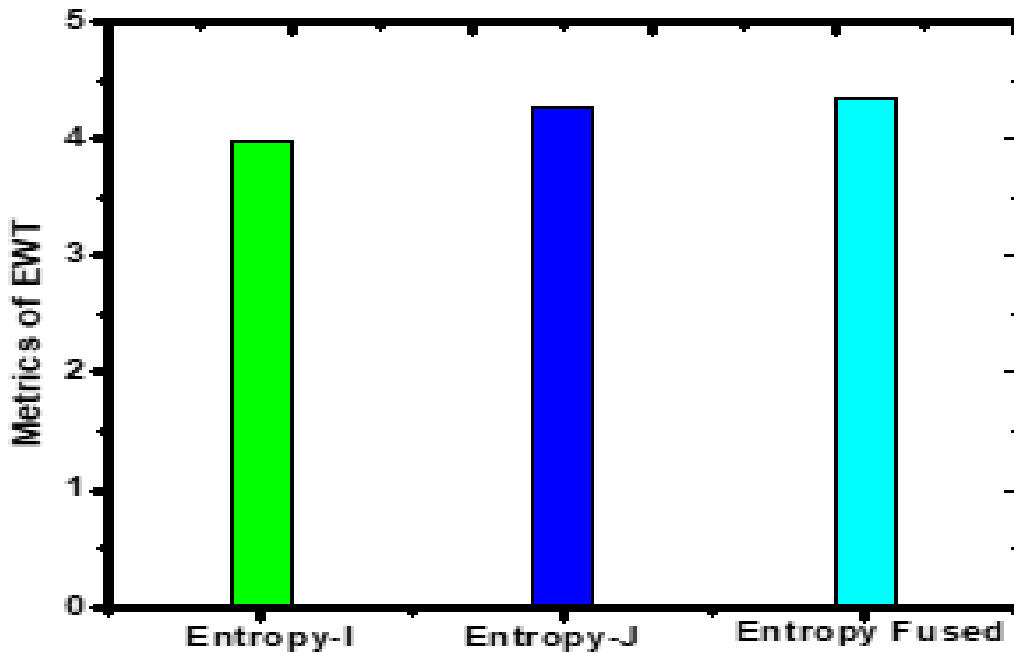


FIGURE 6. Metric of EWT

4. CONCLUSION

In this paper, we proposed a useful method for creating wavelets that are adapted to represent the processed signals. After demonstrating that a tight frame set of wavelets can be created, we defined the Empirical Wavelet Transform and its inverse. The major objective is to identify the illnesses, thoroughly evaluate the organs and tissues, and immediately deliver the diagnosis so that patients can receive immediate care. This study suggests a brand-new fusion technique based on the Empirical Wavelet transform. We combined the photos using the Empirical Wavelet Transform and the fusion rule averages. Standard multi-focus image datasets are used in the experiment. The well-known quality measures are used to assess the efficacy of the fusion processes. In comparison to previous fusion methods, the Empirical Wavelet Transform executes the fusion based on the resolution coefficient, which results in less noise or distortion in the fused pictures. The experimental research also demonstrates that the EWT-based fusion method outperforms the current method.

REFERENCES

- [1]. Cyril Prasanna Raj. P, Venkateshappa, Sudha B S.: High-speed low power DWT architecture for image fusion. International Conference on Electrical, Electronics, and Communications-ICEEC. pp. 156-158 (2014).
- [2]. Heba Khudhair Abass.: A Study of digital image fusion techniques based on contrast and correlation measures, Ph.D. Thesis, Al-Mustansiriyah University (2013).
- [3]. Rani, K., Sharma, R.: Study of Different Image fusion Algorithm. International Journal of Emerging Technology and Advanced Engineering,3(5) pp. 288-291 (2013).
- [4]. Strait,M.,Rahmani,S.,Merkurev,D.: Evaluation of pan-sharpening methods. UCLA Department of Mathematics (2008).
- [5]. Rajvi Patel, Manali Rajput, Prमित Parekh.: Comparative Study on Multi-focus Image Fusion Techniques in Dynamic Scene. International Journal of Computer Applications (2015).
- [6]. Rahmani, S.,Strait, M.,Merkurjev, D., Moeller, M.,Wittman, T.: An adaptive IHS pan sharpening method. Geoscience and Remote Sensing Letters, IEEE, 7(4) pp. 746-750 (2010).
- [7]. Huang, S. G.: Wavelet for image fusion. Graduate Institute of Communication Engineering and Department of Electrical Engineering, National Taiwan University (2010).
- [8]. Naidu, V. P. S.: Image fusion technique using multi-resolution singular value decomposition. Defence Science Journal, 61(5) pp. 479- 484 (2011).
- [9]. Gilles, J.: Empirical Wavelet Transform. Signal Processing, IEEE Transactions on, 61(16) pp. 3999-4010 (2013).
- [10]. Gilles, J., Tran, G.,Osher, S.: 2D Empirical transforms. Wavelets, ridgelets, and curvelets revisited. SIAM Journal on Imaging Sciences, 7(1) pp. 157-186 (2014).
- [11]. Sahu, D. K., Parsai, M. P.: Different image fusion techniques a critical review. International Journal of Modern Engineering Research (IJMER), 2(5) pp. 4298-4301 (2012).
- [12]. N. N. Reddy and D. K. Panda, "A Comprehensive Review on Tunnel Field-Effect Transistor (TFET) Based Biosensors: Recent Advances and Future Prospects on Device Structure and Sensitivity," Silicon, 2020. <https://doi.org/10.1007/s12633-020-00657-1>
- [13]. N. N. Reddy and D. K. Panda, "Simulation Study of Dielectric Modulated Dual Material Gate TFET Based Biosensor by Considering Ambipolar Conduction," Silicon, 2020. DOI:<https://doi.org/10.1007/s12633-020-00784-9>
- [14]. Reddy, N. N., & Panda, D. K. (2021). Performance analysis of Z-shaped gate dielectric modulated (DM) tunnel field-effect transistor- (TFET) based biosensor with extended horizontal N+ pocket. International Journal of Numerical Modelling: Electronic Networks, Devices and Fields, December 2020, 1–13. <https://doi.org/10.1002/jnm.2908>.
- [15]. Reddy, N. N., & Kumar, D. (2021). Nanowire gate all around - TFET - based biosensor by onsidering ambipolar transport. Applied Physics A. <https://doi.org/10.1007/s00339-021-04840-y>.
- [16]. N. Nagendra, D. Kumar, and R. Saha, "International Journal of Electronics and Communications Analytical modelling for surface potential of dual material gate overlapped-on-drain TFET (DM-DMG-TFET) for label-free biosensing application," AEUE - Int. J. Electron. Commun., vol. 151, p. 154225, 2022, doi: 10.1016/j.aeue.2022.154225.
- [17]. N. N. Reddy and D. K. Panda, "Dielectric Modulated Double Gate Hetero Dielectric TFET (DM-DGH-TFET) Biosensors: Gate Misalignment Analysis on Sensitivity," 2022 2nd International Conference on Artificial Intelligence and Signal Processing (AISP), 2022, pp. 1-8, doi: 10.1109/AISP53593.2022.9760561.