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Publications based on thesis work

1. **Journal**

   - **Baruah, H. G.**, Nath, V. K.,and Hazarika, D. (2022) A Remote Sensing Image Retrieval Based on 3D-Local Ternary Pattern (LTP) features and Non-subsampled Shearlet Transform (NSST) Domain Statistical Features, **Computer Modeling in Engineering and Sciences**, 131(1), 137-164(SCIE)
   - **Baruah, H. G.**, Nath, V. K., Hazarika, D., and Hatibaruah, R. (2022). Local bit-plane neighbour dissimilarity pattern in non-subsampled shearlet transform domain for bio-medical image retrieval. **Mathematical Biosciences and Engineering**, 19(2), 1609-1632(SCIE).
   - **Baruah, H.G.**, Nath V.K. and Hazarika, D., CT image retrieval via blend of zernike moment based global shape features and non-subsampled shearlet transform (NSST) domain local texture features. (**Manuscript under preparation**)
   - **Baruah, H.G.**, Nath V.K. and Hazarika, D., Biomedical image retrieval using ZM and SVD based statistical modeling in NSST domain. (**Manuscript under preparation**)

2. **Book Chapter**

   - **Baruah, H. G.**, Nath, V. K., and Hazarika, D. (2019). Remote sensing image retrieval via symmetric normal inverse Gaussian modeling of nonsubsampled shearlet transform coefficients. **In International Conference on Pattern Recognition and Machine Intelligence** (pp. 359-368). Springer, Cham.