

Figure 10. Cross correlation with PAN

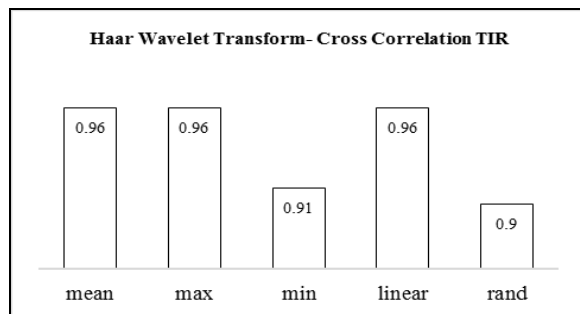


Figure 11. Cross correlation with TIR

7. CONCLUSION

In this paper we evaluated performance of wavelet based image fusion with different filters (mean, linear, max, min and rand) for satellite images (Landsat8 data). Two different approaches used for accuracy assessment of fused images: 1) qualitative approach, 2) quantitative approach.

We showed the performance of different filters in wavelet based image fusion for satellite images. For this purpose, quantitative and qualitative results were considered. It is concluded that linear and mean filters have makes better fused image with respect to visual quality and quantitative values. And also these filters have more stable variation values in different quantitative measures so that they are in first three ranks at all.

Statistical parameters cannot be used for accuracy assessment in fusion process alone. For thermal and visible image fusion accuracy assessment, all parameters (quantitative and qualitative) must be analysed with respect to each other. Qualitative validation is better than quantitative ones. Among all relevant statistical factors, correlation has most meaningful result and similarity to the qualitative assessment. Linear and mean filters have same performance and there is not any difference between their qualitative and quantitative results.

8. REFERENCES

Adelson, E., & Anderson, C. (1984). Pyramid methods in image processing. *RCA Engineer*, 29(6), 33–41. <http://doi.org/10.1.1.59.9419>

Azriel, R., & Mark, T. (1971). Edge and Curve Detection for Visual Scene Analysis. *IEEE Transactions on Computers*, C-20(5), 562–569. <http://doi.org/10.1109/T-C.1971.223290>

Burt P., & Adelson, E. (1983). The Laplacian pyramid as a compact image code. *IEEE Trans. Commun.*, 31(4)(4), 532–540.

Chen, Y., Xiong, J., Liu, H., & Fan, Q. (2014). Optik Fusion method of infrared and visible images based on neighborhood

characteristic and regionalization in NSCT domain. *Optik - International Journal for Light and Electron Optics*, 125(17), 4980–4984. <http://doi.org/10.1016/j.ijleo.2014.04.006>

Christine Pohl, J. van G. (2017). *Remote Sensing Image Fusion A Practical Guide*. CRC Press.

Cui, G., Feng, H., Xu, Z., Li, Q., & Chen, Y. (2015). Detail preserved fusion of visible and infrared images using regional saliency extraction and multi-scale image decomposition, 341, 199–209. <http://doi.org/10.1016/j.optcom.2014.12.032>

Florack, L. M. J., ter Haar Romeny, B. M., Koenderink, J. J., & Viergever, M. A. (1994). Linear scale-space. *Journal of Mathematical Imaging and Vision*, 4(4), 325–351. <http://doi.org/10.1007/BF01262401>

Kun, L., Lei, G., Huihui, L., & Jingsong, C. (2009). Fusion of Infrared and Visible Light Images Based on Region Segmentation. *Chinese Journal of Aeronautics*, 22(1), 75–80. [http://doi.org/10.1016/S1000-9361\(08\)60071-0](http://doi.org/10.1016/S1000-9361(08)60071-0)

Luis, G., Member, S., Tuia, D., & Member, S. (2015). Multimodal Classification of Remote Sensing Images: A Review and Future Directions, 1–52. <http://doi.org/10.1109/JPROC.2015.2449668>

Misiti, M., & Poggi, J. (n.d.). Wavelet Toolbox For Use with MATLAB.

Morales, E., & Shih, F. Y. (2000). Wavelet coefficients clustering using morphological operations and pruned quadrees, 33, 1611–1620. [http://doi.org/10.1016/S0031-3203\(99\)00147-8](http://doi.org/10.1016/S0031-3203(99)00147-8)

Pajares, G., & de la Cruz, J. M. (2004). A wavelet-based image fusion tutorial. *Pattern Recognition*, 37(9), 1855–1872. <http://doi.org/10.1016/j.patcog.2004.03.010>

Schowengerdt Robert A. (2009). *Remote Sensing: Models And Methods For Image Processing, 3Rd Edition*. Elsevier Science & Technology (2009).

Shi, W., Zhu, C., Tian, Y., & Nichol, J. (2005). Wavelet-based image fusion and quality assessment. *International Journal of Applied Earth Observation and Geoinformation*, 6(3–4), 241–251. <http://doi.org/10.1016/j.jag.2004.10.010>

Stollnitz, E. J., DeRose, a. D., & Salesin, D. H. (1995). Wavelets for computer graphics: a primer.1. *IEEE Computer Graphics and Applications*, 15(May), 1–8. <http://doi.org/10.1109/38.376616>

Transactions, I., Pattern, O. N., & Vol, M. I. (1989). A Theory for Multiresolution Signal Decomposition: The Wavelet Representation, 1(7), 674–693.

Vivone, G., Alparone, L., Chanussot, J., Mura, M. D., Garzelli, A., Member, S., ... Wald, L. (2014). Pansharpening Algorithms. *IEEE Transactions on Geoscience and Remote Sensing*, 53(5), 2565–2586.

Vivone, G., Simões, M., Mura, M. D., Restaino, R., Bioucas-dias, J. M., Licciardi, G. A., & Chanussot, J. (2014). Pansharpening Based on Semiblind Deconvolution, 1–14.

Wald, L. (2009). Some terms of reference in data fusion SOME TERMS OF REFERENCE IN DATA, 37(3), 1190–1193.

Wang, Z., Ziou, D., Armenakis, C., Li, D., & Li, Q. (2005). A Comparative Analysis of Image Fusion Methods, 43(6), 1391–1402.

Zhang, B., Lu, X., & Jia, W. (2013). Optik A multi-focus image fusion algorithm based on an improved dual-channel PCNN in NSCT domain. *Optik - International Journal for Light and Electron Optics*, 124(20), 4104–4109. <http://doi.org/10.1016/j.ijleo.2012.12.032>

Zhao, J., Feng, H., Xu, Z., Li, Q., & Liu, T. (2013). Detail enhanced multi-source fusion using visual weight map extraction based on multi scale edge preserving decomposition, 287, 45–52. <http://doi.org/10.1016/j.optcom.2012.08.070>