

A High-Efficiency Satellite Image Classification Method for Real-Time Application Using Augmented Incremental Transfer Learning

Dr. Akshay Dhande

Department of Electronics and Telecommunication Engineering,
Lovely Professional University, Punjab, India
akshaydhande126@gmail.com,
<https://orcid.org/0000-0002-7340-9806>

Abstract: Researchers have put forth a range of models for processing satellite images, each with distinct data formats and processing requirements. One limitation arises from variations in the forms of modules, such as the image capturing and feature extraction modules, impacting accuracy and scalability in real-time scenarios. This passage introduces and explores a pioneering very competent temporal engine crafted specifically for classification of real-time satellite images. The proposed approach utilizes augmented incremental transfer learning, aiming to mitigate the limitations associated with diverse processing requirements. This approach involves capturing real-time satellite data through Google's Earth Engine and subsequently processing it using a Convolutional Neural Network (CNN) based on transfer learning. The CNN employs backscatter coefficient analysis, utilizing coefficients derived from Precision Image's average intensity value across a distributed target. By integrating incremental learning and CNN for classification, the model achieves an impressive average accuracy of 98.06% in detecting crop type and severity of damage. Comparative analysis with state-of-the-art approaches reveals the superiority of the proposed model. It outperforms existing models by 5% in accuracy, showcasing its efficacy in satellite image processing and classification.

Keywords: Earth, Satellite, crop, type, damage, classification, deep learning, incremental, transfer, accuracy, precision, recall

REFERENCES

- [1] R. Luciani, G. Laneve and M. JahJah, "Agricultural Monitoring, an Automatic Procedure for Crop Mapping and Yield Estimation: The Great Rift Valley of Kenya Case," in IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing, vol. 12, no. 7, pp. 2196-2208, July 2019, doi: 10.1109/JSTARS.2019.2921437.
- [2] A. Shelestov et al., "Cloud Approach to Automated Crop Classification Using Sentinel-1 Imagery," in IEEE Transactions on Big Data, vol. 6, no. 3, pp. 572-582, 1 Sept. 2020,
- [3] J. Jiang et al., "HISTIF: A New Spatiotemporal Image Fusion Method for High-Resolution Monitoring of Crops at the Subfield Level," in IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing, vol. 13, pp. 4607-4626, 2020,
- [4] P. Tang, P. Du, J. Xia, P. Zhang and W. Zhang, "Channel Attention-Based Temporal Convolutional Network for Satellite Image Time Series Classification," in IEEE Geoscience and Remote Sensing Letters, vol. 19, pp. 1-5, 2022, Art no. 8016505,
- [5] M. H. Asad and A. Bais, "Crop and Weed Leaf Area Index Mapping Using Multi-Source Remote and Proximal Sensing," in IEEE Access, vol. 8, pp. 138179-138190, 2020,
- [6] S. Liu, Z. Zhou, H. Ding, Y. Zhong and Q. Shi, "Crop Mapping Using Sentinel Full-Year Dual-Polarized SAR Data and a CPU-Optimized Convolutional Neural Network with Two Sampling Strategies," in IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing, vol. 14, pp. 7017-7031, 2021, doi: 10.1109/JSTARS.2021.3094973.

- [7] M. M. G. de Macedo, A. B. Mattos and D. A. B. Oliveira, "Generalization of Convolutional LSTM Models for Crop Area Estimation," in IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing, vol. 13, pp. 1134-1142, 2020,
- [8] C. Silva-Perez, A. Marino, J. M. Lopez-Sanchez and I. Cameron, "Multitemporal Polarimetric SAR Change Detection for Crop Monitoring and Crop Type Classification," in IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing, vol. 14, pp. 12361-12374, 2021,
- [9] S. Yang, L. Gu, X. Li, F. Gao and T. Jiang, "Fully Automated Classification Method for Crops Based on Spatiotemporal Deep-Learning Fusion Technology," in IEEE Transactions on Geoscience and Remote Sensing, vol. 60, pp. 1-16, 2022, Art no. 5405016,
- [10] H. -W. Jo et al., "Deep Learning Applications on Multitemporal SAR (Sentinel-1) Image Classification Using Confined Labeled Data: The Case of Detecting Rice Paddy in South Korea," in IEEE Transactions on Geoscience and Remote Sensing, vol. 58, no. 11, pp. 7589-7601, Nov. 2020, doi: 10.1109/TGRS.2020.2981671.
- [11] T. Lampert, B. Lafabregue, T. -B. -H. Dao, N. Serrette, C. Vrain and P. Gañarski, "Constrained Distance-Based Clustering for Satellite Image Time-Series," in IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing, vol. 12, no. 11, pp. 4606-4621, Nov. 2019, doi: 10.1109/JSTARS.2019.2950406.
- [12] J. Bell, E. Gebremichael, A. Molthan, L. Schultz, F. Meyer and S. Shrestha, "Synthetic Aperture Radar and Optical Remote Sensing of Crop Damage Attributed to Severe Weather in the Central United States," IGARSS 2019 - 2019 IEEE International Geoscience and Remote Sensing Symposium, 2019, pp. 9938-9941, doi: 10.1109/IGARSS.2019.8899775.
- [13] S. Jones and J. Saniie, "Using Deep Learning and Satellite Imagery to Assess the Damage to Civil Structures After Natural Disasters," 2019 IEEE International Conference on Electro Information Technology (EIT), 2019, pp. 189-193, doi: 10.1109/EIT.2019.8833724.
- [14] Y. Sofue, C. Hongo, N. Manago, G. Sigit, K. Homma and B. Barus, "Estimation of Normal Rice Yield Considering Heading Stage Based on Observation Data and Satellite Imagery," 2021 IEEE International Geoscience and Remote Sensing Symposium IGARSS, 2021, pp. 6439-6442, doi: 10.1109/IGARSS47720.2021.9554679.
- [15] D. Lakmal, K. Kugathasan, V. Nanayakkara, S. Jayasena, A. S. Perera and L. Fernando, "Brown Planthopper Damage Detection using Remote Sensing and Machine Learning," 2019 18th IEEE International Conference On Machine Learning And Applications (ICMLA), 2019
- [16] Z. Li, G. Chen and T. Zhang, "A CNN-Transformer Hybrid Approach for Crop Classification Using Multitemporal Multisensor Images," in IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing, vol. 13, pp. 847-858, 2020, doi: 10.1109/JSTARS.2020.2971763.