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A Practical Study on Banana (Musa spp.) Plant Counting and Coverage Percentage Using Remote Sensing and Deep Learning

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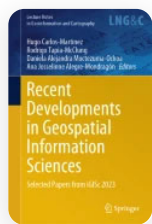
Recent Developments in Geospatial Information Sciences

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Abstract

This paper focuses on a practical study to count banana plants and evaluate their coverage in RGB images. Two approaches were used: GLI and K-means algorithms and the YOLOv5m model. The GLI and K-means combination demonstrated the ability to identify the center of the banana plant. The YOLOv5m deep learning approach showed solid performance, evidenced by a significant 86% precision and 90% recall during training. Overall, the methods showed favorable results in the evaluation, with an IoU of 81% for GLI and K-means and 76% for the YOLOv5m application. Regarding coverage estimation, the U-Net architecture achieved an IoU of 52.3% in the testing data. However, the model sought to adapt to the morphology of banana plants. This study contributes to the informed selection of techniques for agricultural monitoring and analysis of crops such as banana.

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References

Aeberli A, Johansen K, Robson A, Lamb DW, Phinn S (2021) Detection of banana plants using multi-temporal multispectral UAV imagery. Remote Sens 13(11):2123. https://doi.org/10.3390/rs13112123

Article   Google Scholar

Aeberli A, Phinn S, Johansen K, Robson A, Lamb DW (2023) Characterisation of banana plant growth using high-spatiotemporal-resolution multispectral UAV imagery. Remote Sens 15(3):679. https://doi.org/10.3390/rs15030679

Article   Google Scholar

Banerjee D, Kukreja V, Hariharan S, Sharma V (2023) Precision agriculture: classifying banana leaf diseases with hybrid deep learning models. In: 2023 IEEE 8th international conference for convergence in technology (I2CT). Lonavla, India. IEEE, pp 1–5. https://doi.org/10.1109/I2CT57861.2023.10126431

Benos L, Tagarakis AC, Dolias G, Berruto R, Kateris D, Bochtis D (2021) Machine learning in agriculture: a comprehensive updated review. Sensors 21(11):3758. https://doi.org/10.3390/s21113758

Article   Google Scholar

Calou VBC, Teixeira ADS, Moreira LCJ, Lima CS, De Oliveira JB, De Oliveira MRR (2020) The use of UAVs in monitoring yellow sigatoka in banana. Biosyst Eng 193:115–125. https://doi.org/10.1016/j.biosystemseng.2020.02.016

Article   Google Scholar

DANE (2023) Enfermedades y plagas del plátano (Musa paradisiaca) y el banano (Musa acuminata; M sapientum) en Colombia. Departamento administrativo nacional de estadística (DANE), Sep 2016. Accessed 31 Aug. https://www.dane.gov.co/files/investigaciones/agropecuario/sipsa/Bol_Insumos_sep_2016.pdf

David et al (2021) Plant detection and counting from high-resolution RGB images acquired from UAVs: comparison between deep-learning and handcrafted methods with application to maize, sugar beet, and sunflower. Plant Biol, preprint. https://doi.org/10.1101/2021.04.27.441631

Du J (2018) Understanding of object detection based on CNN family and YOLO. J Phys Conf Ser 1004:012029. https://doi.org/10.1088/1742-6596/1004/1/012029

Article   Google Scholar

Eng LS, Ismail R, Hashim W, Baharum A (2019) The use of VARI, GLI, and VIgreen formulas in detecting vegetation in aerial images. Int J Technol 10(7):1385. https://doi.org/10.14716/ijtech.v10i7.3275

Article   Google Scholar

Fan X, Yan C, Fan J, Wang N (2022) Improved U-net remote sensing classification algorithm fusing attention and multiscale features. Remote Sens 14(15):3591. https://doi.org/10.3390/rs14153591

Article   Google Scholar

Gadiraju KK, Vatsavai RR (2020) Comparative analysis of deep transfer learning performance on crop classification. In: Proceedings of the 9th ACM SIGSPATIAL international workshop on analytics for big geospatial data. Seattle Washington, ACM, pp 1–8. https://doi.org/10.1145/3423336.3431369

Gomez Selvaraj M et al (2020) Detection of banana plants and their major diseases through aerial images and machine learning methods: a case study in DR Congo and Republic of Benin. ISPRS J Photogramm Remote Sens 169:110–124. https://doi.org/10.1016/j.isprsjprs.2020.08.025

Joshi A, Pradhan B, Gite S, Chakraborty S (2023) Remote-sensing data and deep-learning techniques in crop mapping and yield prediction: a systematic review. Remote Sens 15(8):2014. https://doi.org/10.3390/rs15082014

Article   Google Scholar

Karthi M, Muthulakshmi V, Priscilla R, Praveen P, Vanisri K (2021) Evolution of YOLO-V5 algorithm for object detection: automated detection of library books and performace validation of dataset. In: 2021 international conference on innovative computing, intelligent communication and smart electrical systems (ICSES). Chennai, India. IEEE, pp 1–6. https://doi.org/10.1109/ICSES52305.2021.9633834

Kilwenge R, Adewopo J, Sun Z, Schut M (2021) UAV-based mapping of banana land area for village-level decision-support in Rwanda. Remote Sens 13(24):4985. https://doi.org/10.3390/rs13244985

Article **Google Scholar**

Lamour J, Naud O, Lechaudel M, Tisseyre B (2017) Mapping properties of an asynchronous crop: the example of time interval between flowering and maturity of banana. Adv Anim Biosci 8(2):481–486. https://doi.org/10.1017/S2040470017000449

Article **Google Scholar**

Louhaichi M, Borman MM, Johnson DE (2001) Spatially located platform and aerial photography for documentation of grazing impacts on wheat. Geocarto Int 16(1):65–70. https://doi.org/10.1080/10106040108542184

Article **Google Scholar**

Novara M, Estergaard S, Olmos D, Rasjido J, Soto A (2019) Comparación de índices espectrales RGB, a partir de imágenes VANT, para la identificación de coberturas vegetales en ambientes urbanos y rurales patagónicos esteparios. Accessed 28 Aug 2023. https://www.researchgate.net/publication/337292177_Comparacion_de_indices_espectrales_RGB_a_partir_de_imagenes_VANT_para_la_identificacion_de_coberturas_vegetales_en_ambientes_urbanos_y_rurales_patagonicos_esteparios

GAD Puebloviejo (2014) Plan de desarrollo y ordenamiento territorial 2015–2025

Google Scholar

Redmon J, Divvala S, Girshick R, Farhadi A (2015) You only look once: unified, real-time object detection. https://doi.org/10.48550/ARXIV.1506.02640

Ribani R, Marengoni M (2019) A survey of transfer learning for convolutional neural networks. In: 2019 32nd SIBGRAPI conference on graphics, patterns and images tutorials (SIBGRAPI-T). Rio de Janeiro, Brazil. IEEE, pp 47–57. https://doi.org/10.1109/SIBGRAPI-T.2019.00010

Ronneberger O, Fischer P, Brox T (2015) U-Net: convolutional networks for biomedical image segmentation. https://doi.org/10.48550/ARXIV.1505.04597

Shirzadifar A, Maharlooei M, Bajwa SG, Oduor PG, Nowatzki JF (2020) Mapping crop stand count and planting uniformity using high resolution imagery in a maize crop. Biosyst Eng 200:377–390. https://doi.org/10.1016/j.biosystemseng.2020.10.013

Article **Google Scholar**

Valdez AE, Castaneda MAP, Gomez-Rojas J, Ramos RL (2022) Canopy extraction in a banana crop from UAV captured multispectral images. In: 2022 IEEE 40th central America and Panama convention (CONCAPAN). Panama, Panama. IEEE, pp 1–6. https://doi.org/10.1109/CONCAPAN48024.2022.9997598

Vit A, Shani G (2018) Comparing RGB-D sensors for close range outdoor agricultural phenotyping. Sensors 18(12):4413. https://doi.org/10.3390/s18124413

Article **Google Scholar**

Voora V, Bermúdez S, Farrell JJ, Larrea C, Luna E (2023) Global market report: banana prices and sustainability

Google Scholar

Wang L, Wang J, Zhang X, Wang L, Qin F (2022) Deep segmentation and classification of complex crops using multi-feature satellite imagery. Comput Electron Agric 200:107249. https://doi.org/10.1016/j.compag.2022.107249

Article **Google Scholar**

Xu Y et al (2021) Sustainability trait modeling of field-grown switchgrass (Panicum virgatum) using UAV-based imagery. Plants 10(12):2726. https://doi.org/10.3390/plants10122726

Article **CAS** **Google Scholar**

Yan C, Fan X, Fan J, Wang N (2022) Improved U-net remote sensing classification algorithm based on multi-feature fusion perception. Remote Sens 14(5):1118. https://doi.org/10.3390/rs14051118

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