

Towards smarter agriculture: Automatic identification of crop heads with artificial intelligence

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The top two rows show extracted image frames from the background video clips—i.e., video clips of fields with no wheat plants. The third row shows examples of image frames extracted from the video clip of a wheat field. Credit: *Plant Phenomics* (2023). DOI: 10.34133/plantphenomics.0025

Recent advances in artificial intelligence (AI), alongside drones and digital cameras, have greatly extended the frontiers of smart agriculture. One attractive use case for these technologies is precision agriculture. In this modern approach to farming, the idea is to optimize crop production by gathering precise data about plants and the state of the field, and then act accordingly.

For example, by analyzing aerial images of crops, AI models can determine what parts of a field need more attention, as well as the current stage of development of the plants.

Among all the crop monitoring functions that AI can do, crop head counting remains as one of the most challenging to implement. Images of crops contain densely packed, repeating patterns that are usually irregular and overlapped, making it difficult for <u>deep learning models</u> to automatically detect specific plant organs. Ideally, one would train such models using thousands of manually annotated images, in which pixels belonging to crop heads are pre-specified. In practice, however, annotating crop images is extremely tedious and time-consuming.

To address this issue, a research team including Assistant Professor Lingling Jin from the University of Saskatchewan, Canada, developed an innovative technique that can simplify the training and development of deep learning models. Their approach, which is described in a paper recently made available in *Plant Phenomics*, could promote a more widespread adoption of AI in agriculture.



To illustrate their idea, the team focused on the identification (or 'segmentation') of <u>wheat</u> heads in crop images as an example use case. Their strategy revolves around generating a synthetic annotated dataset. That is, instead of manually marking pixels belonging to wheat heads in hundreds of images, they devised a convenient way to produce artificial images in which the wheat heads are automatically marked.

First, the researchers recorded short videos of a wheat field and of other locations without wheat plants (also called 'background' videos). From the footage of the wheat field, they extracted a small number of still frames and manually annotated them, identifying all the wheat heads.

Then, using frames from the background videos as a canvas, they generated synthetic wheat images by pasting 'cutouts' of the manually segmented wheat heads onto them. This approach enabled the team to produce thousands of training images for a deep learning model with minimal effort.

To further improve the model, which was based on a customized U-Net architecture, the researchers also employed various domain adaptation techniques. These techniques fine-tuned the algorithm so that it would perform better on images from various real-world wheat fields, even though it was trained mainly on synthetic images.

Numerous tests on an open-access dataset revealed impressive gains in accuracy, as Jin notes, "Our approach established—and by a wide margin in performance—a new state-of-the-art model for wheat head segmentation."

Worth noting, the techniques showcased in this work are not limited to identifying wheat heads. In this regard, Jin says, "While we showed the utility of the proposed method for wheat head segmentation, it could be applied to other applications that have similar dense repeating patterns of



objects, such as segmenting plant organs in other crop species or segmenting molecular components in microscopy images." Hence, this work paints a bright future for deep learning both in agriculture and other fields.

More information: Keyhan Najafian et al, Semi-Self-Supervised Learning for Semantic Segmentation in Images with Dense Patterns, *Plant Phenomics* (2023). <u>DOI: 10.34133/plantphenomics.0025</u>

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