

Vision-based automated systems are a promising way to speed up the identification process of pest psyllids like D. citri and T. erytreae

## **Potential for automation of citrus pest identification using computer vision-based** artificial intelligence recognition

| Background:  | Psyllid pests are significant vectors of a greening disease that affects citrus. <i>Trioza erytreae</i> , a native psyllid, transmits citrus greening disease associated with ' <i>Candidatus</i> Liberibacter africanus'   |           |        |                            |                 |
|--|---|-----------|--------|----------------------------|-----------------|
|  | However, South Africa faces the threat of an incursion by <i>Diaphorina citri</i> , a highly efficient vector of Huanglongbing (HLB) that is associated with ' <i>Candidatus</i> Liberibacter asiaticus' Detection of psyllid pests is crucial for effective management, with yellow sticky traps being commonly used |           |        |                            |                 |
|  |   |           |        |                            |                 |
|  | Identification of psyllids on traps is done manually, but is time consuming and delays control needed to prevent disease spread   |           |        |                            |                 |
| Result:  | Mean average precision (mAP), precision, recall and time per sticky trap for five trained models.   |           |        |                            |                 |
| YOLOv8s was most precise, but YOLOv8m was the most sensitive without sacrificing accuracy. |   |           |        |                            |                 |
| Model  | mAP   | Precision | Recall | Time per trap<br>(seconds) |                 |
| YOLOv8s  | 0.849   | 0.905     | 0.777  | 6.7                        |                 |
| YOLOv8m  | 0.845   | 0.897     | 0.797  | 12.7                       |                 |
| YOLOv8n  | 0.831   | 0.742     | 0.859  | 3.7                        |                 |
| YOLOv8l  | 0.797   | 0.797     | 0.786  | 21.4                       |                 |
| YOLOv8x  | 0.781   | 0.777     | 0.689  | 29.2                       | Trioza erytreae |
| Methods  |   |           |        |                            |                 |



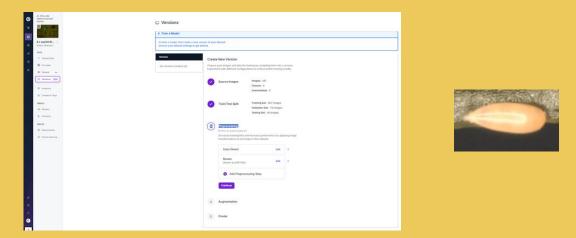
## **NIELIIOUS**

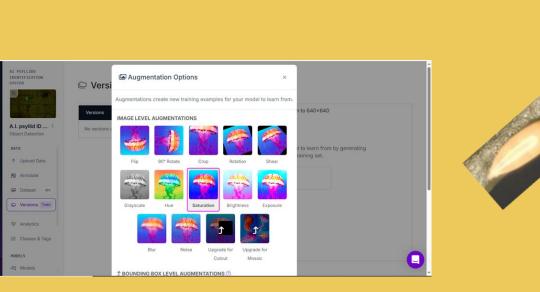


Trap collection



## Trap photography





12 x 7 \* \* 0 \*

Image annotation

The A.I. developmental process from trap collection to model testing, using 875 cropped images that were augmented, processed, and then split into training, validation, and testing datasets in a ratio of 7:2:1

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