



# FlowerPhenoNet: Automated Flower Detection from Multi-view Image Sequences using Deep Neural Networks for Temporal Plant Phenotyping Analysis



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## FlowerPhenoNet: Novelties

- ❖ A novel approach to flower detection from a multi-view image sequence using deep learning for plant phenotyping.
- ❖ A set of new temporal flower phenotypes and their significance in plant science.
- ❖ A publicly available benchmark dataset to facilitate research advancement in flower-based plant phenotyping.

## FlowerPhenoNet: Steps

- ❖ *Image labeling*: The flowers in the training set are manually enclosed by rectangular boxes.
- ❖ *Data augmentation*: Application of geometric and photometric transformation to the labelled training set to increase its size.
- ❖ *Neural network architecture and training*: You Only Look Once (YOLO) object detector is used for flower detection.
- ❖ *Testing*: Image sequences consisting of images of all days for available views are used to test the performance of the detector.
- ❖ *Phenotype computation*: Computed phenotypes fall in two broad categories-
  - *Trajectory-based*: Denoted by graphical representation of  $[p_1, p_2, \dots, p_n]$ , where  $p_i$  represents flower size (for flower growth trajectory) or total flower count (for blooming trajectory) for the  $i$ -th image of the plant.
  - *Event-based*: Timing of emergence of the first flower, the total number of flowers present at any time in the image sequence, and the size of each flower at any time.

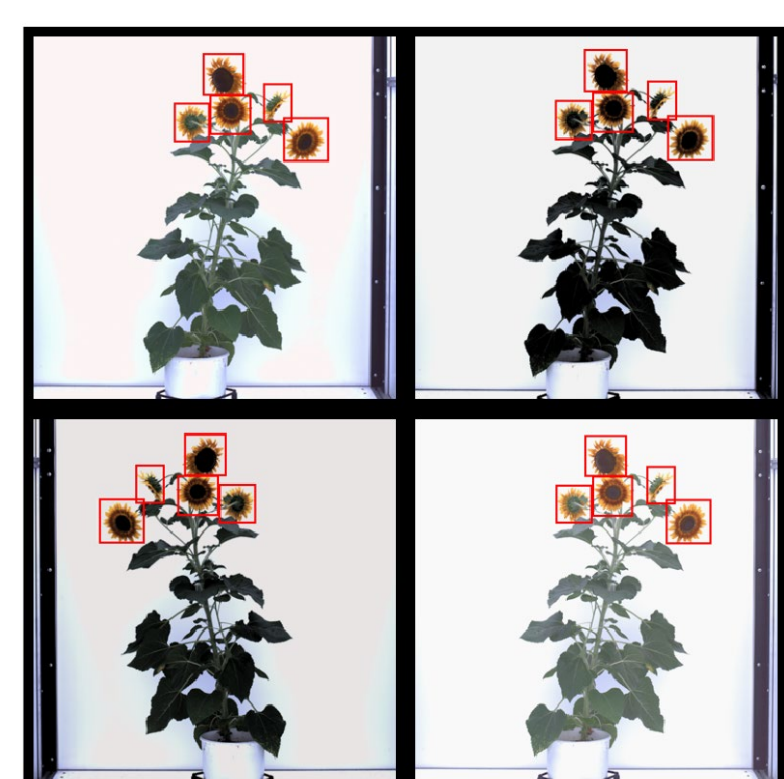


Fig 2: Data augmentation

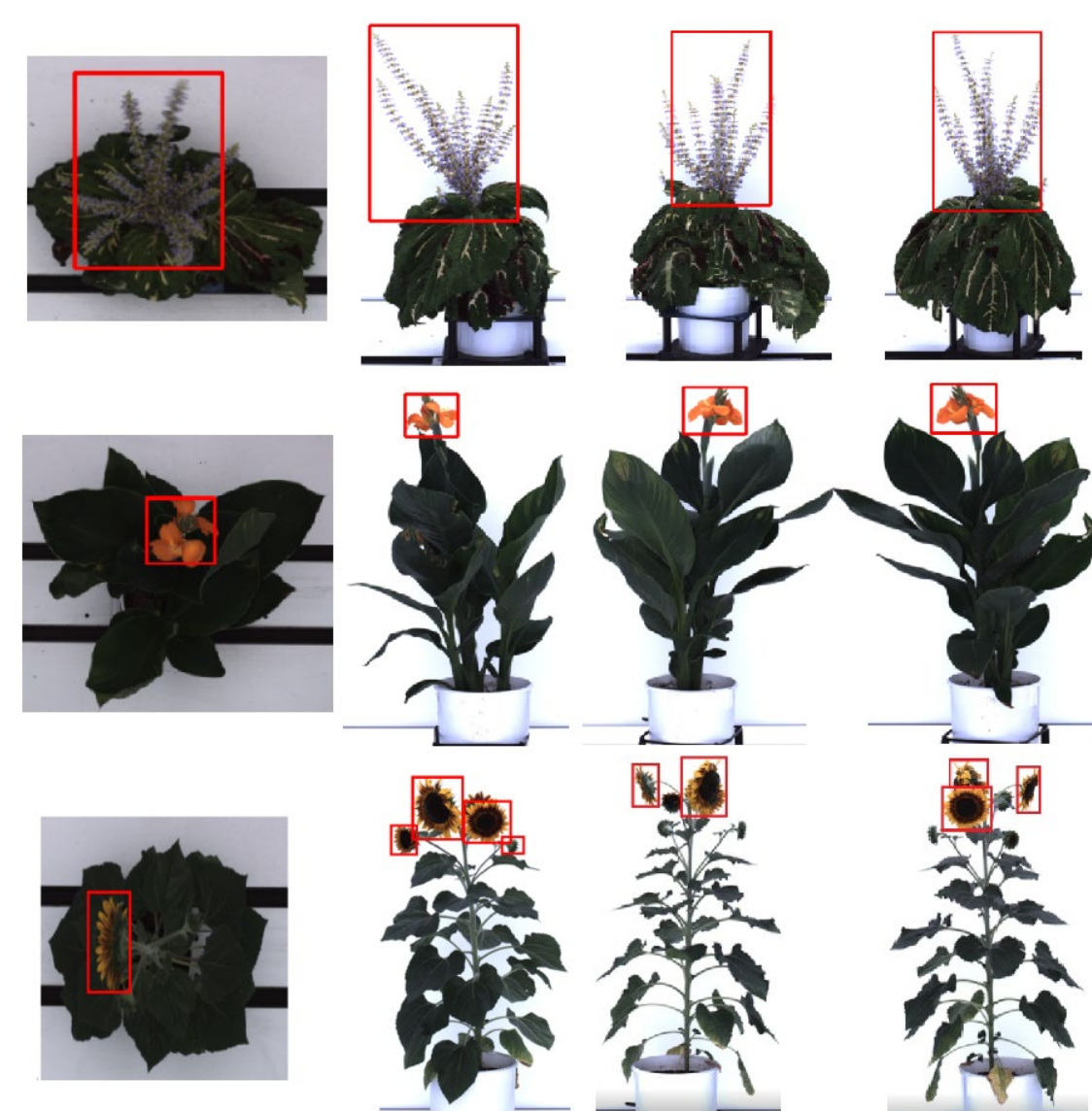


Fig 3: Sample labelled images from FlowerPheno dataset

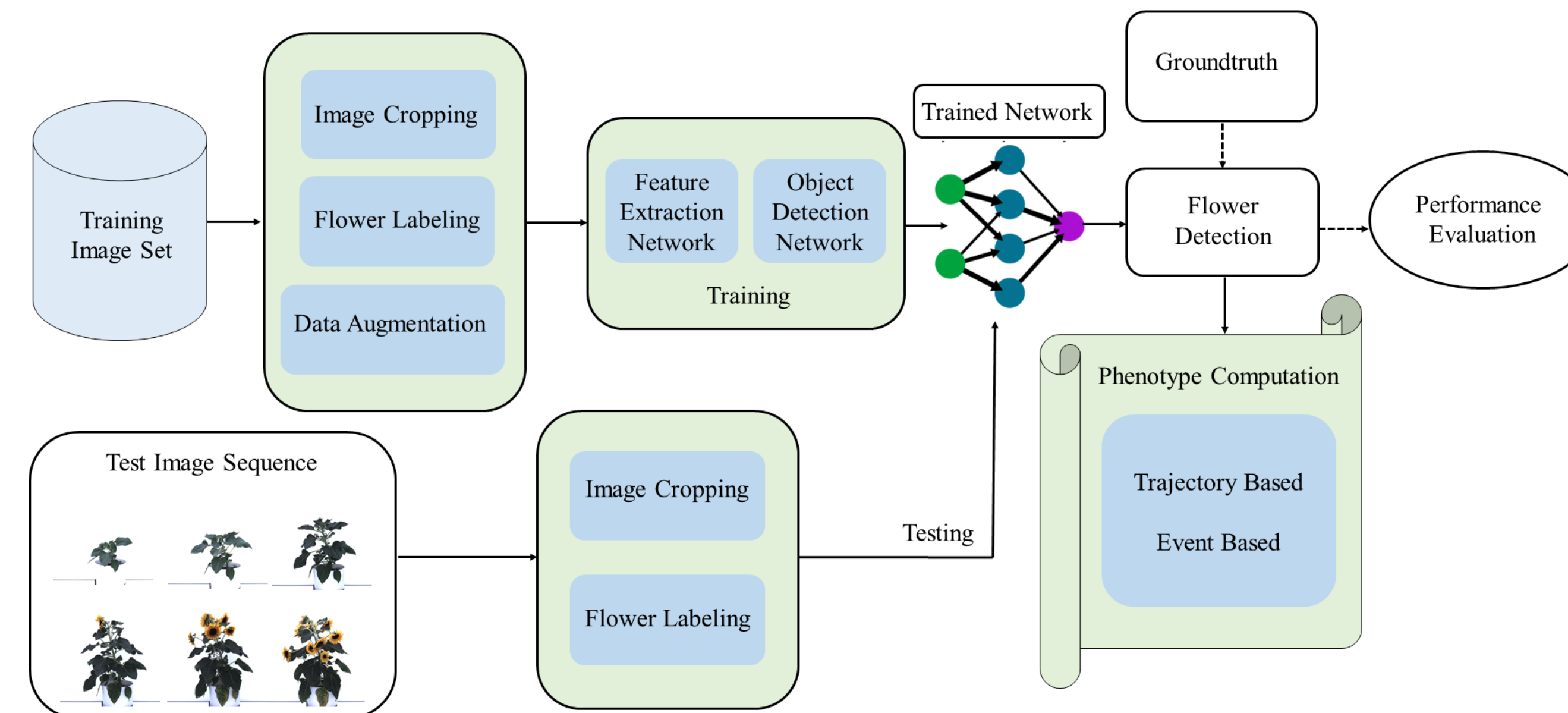


Fig. 1: Block diagram of FlowerPhenoNet

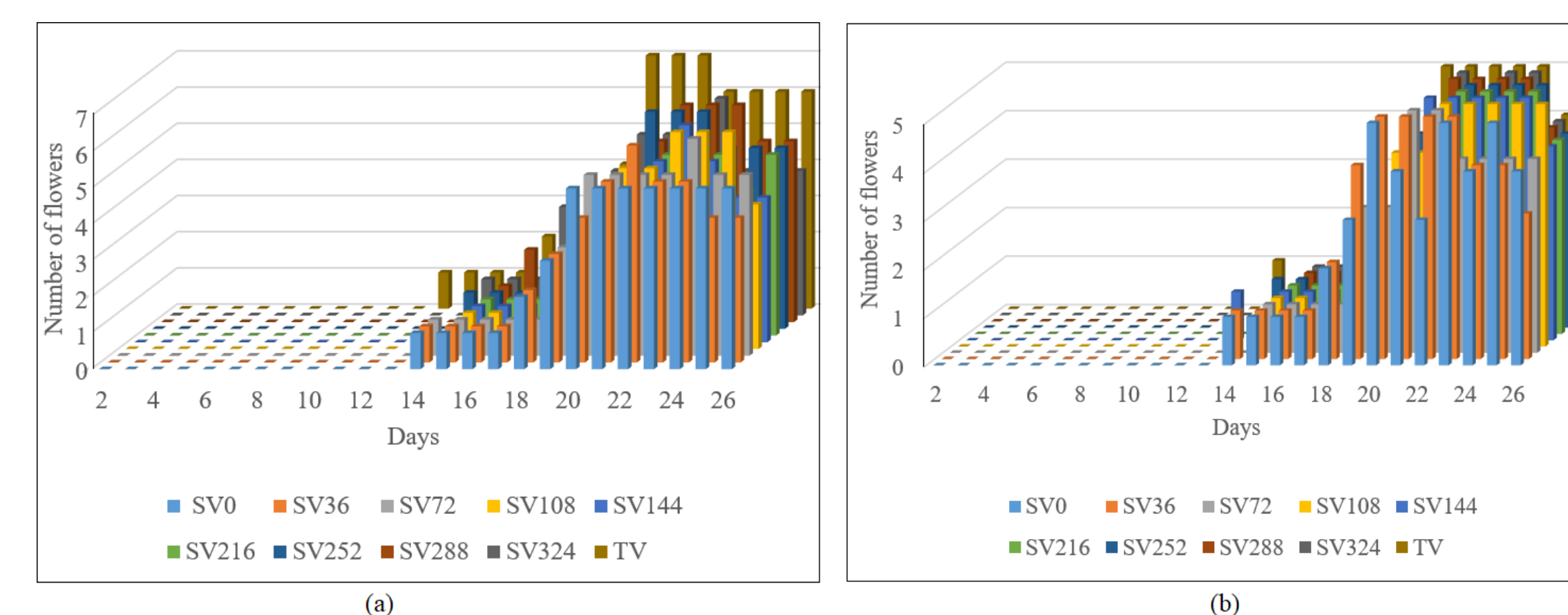


Fig. 4: The flower status graphs for two sample sunflower plants

Flower status report provides the following important information:

- ❖ The emergence day of the flower in each view
- ❖ The total number of flowers present for a given view on any day
- ❖ The highest number of flowers bloomed in the plant

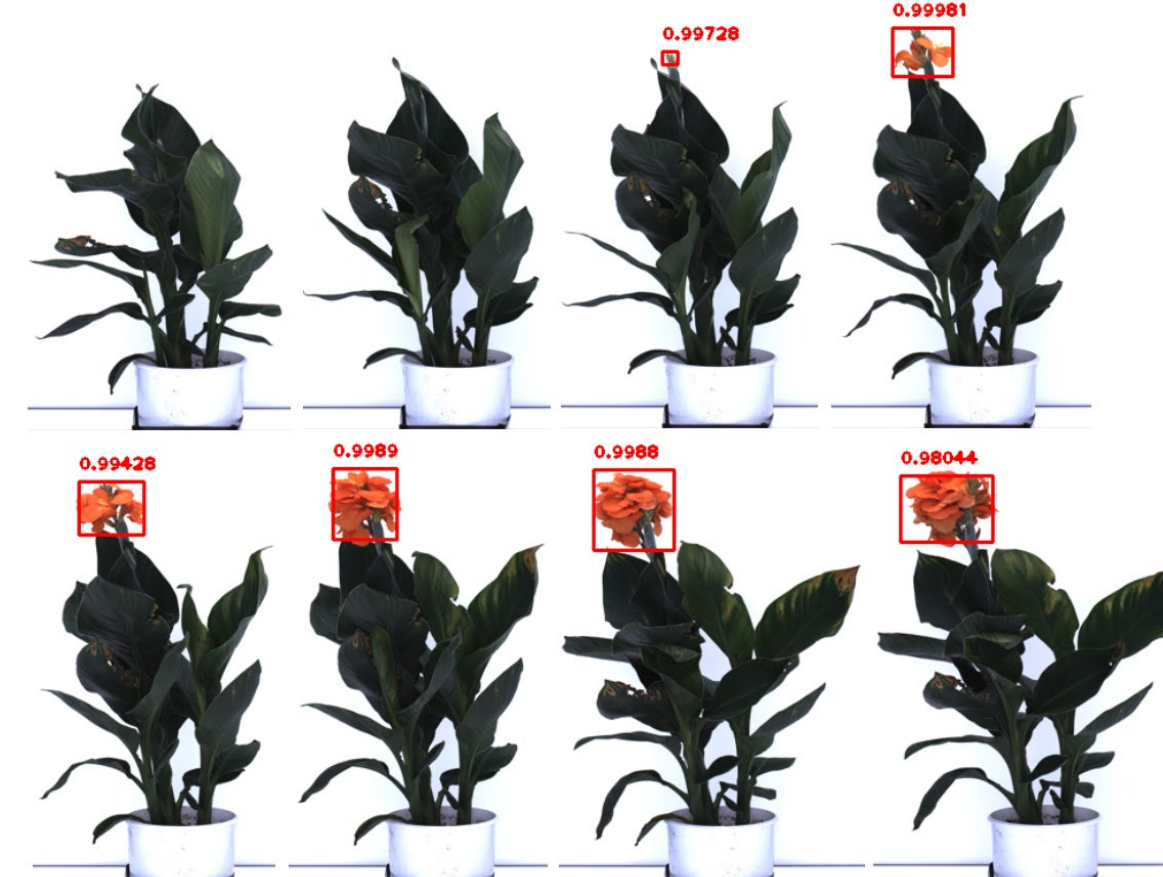


Fig 5: Illustration of growth of canna flower over time

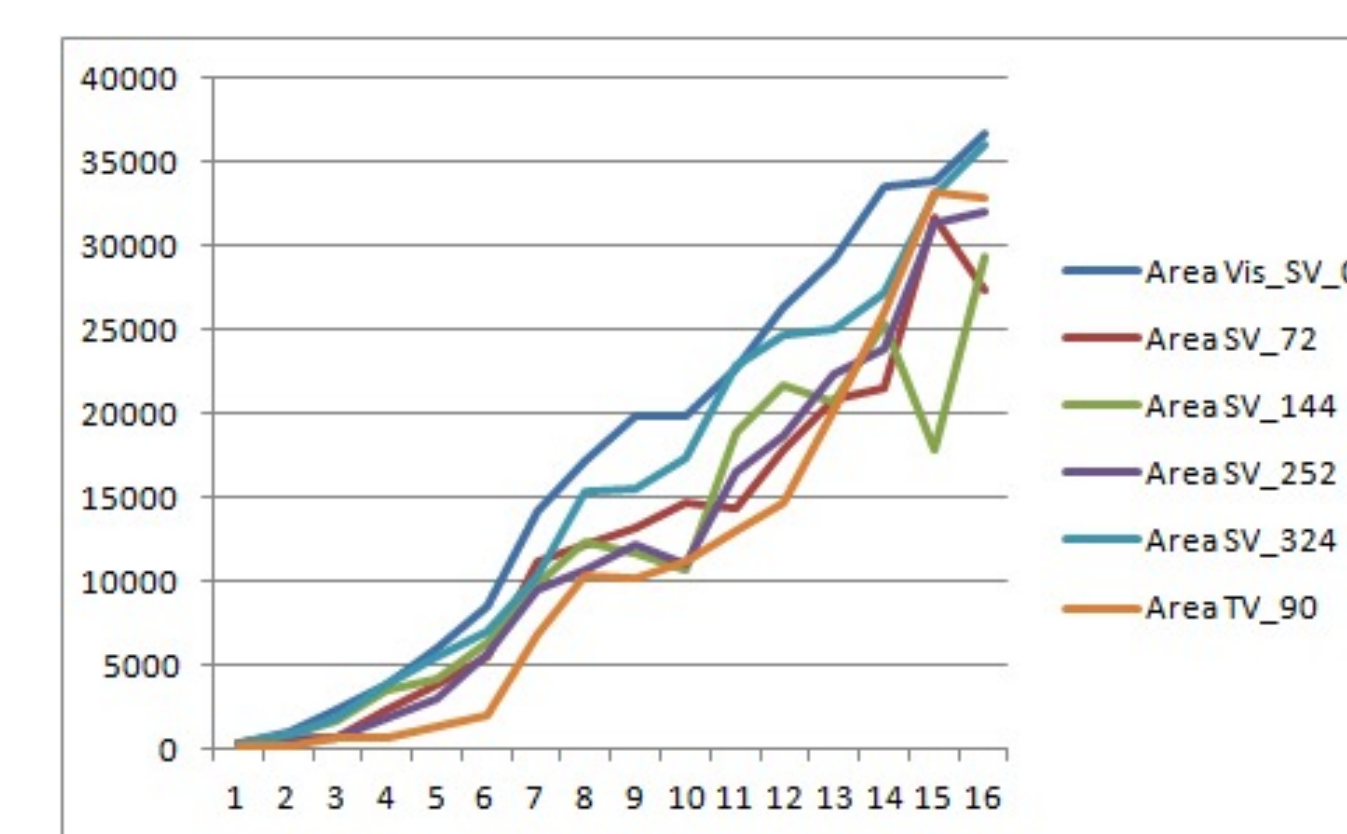


Fig 6: Growth trajectory of a canna flower in multiple views

## Acknowledgement

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## Temporal Phenotypes

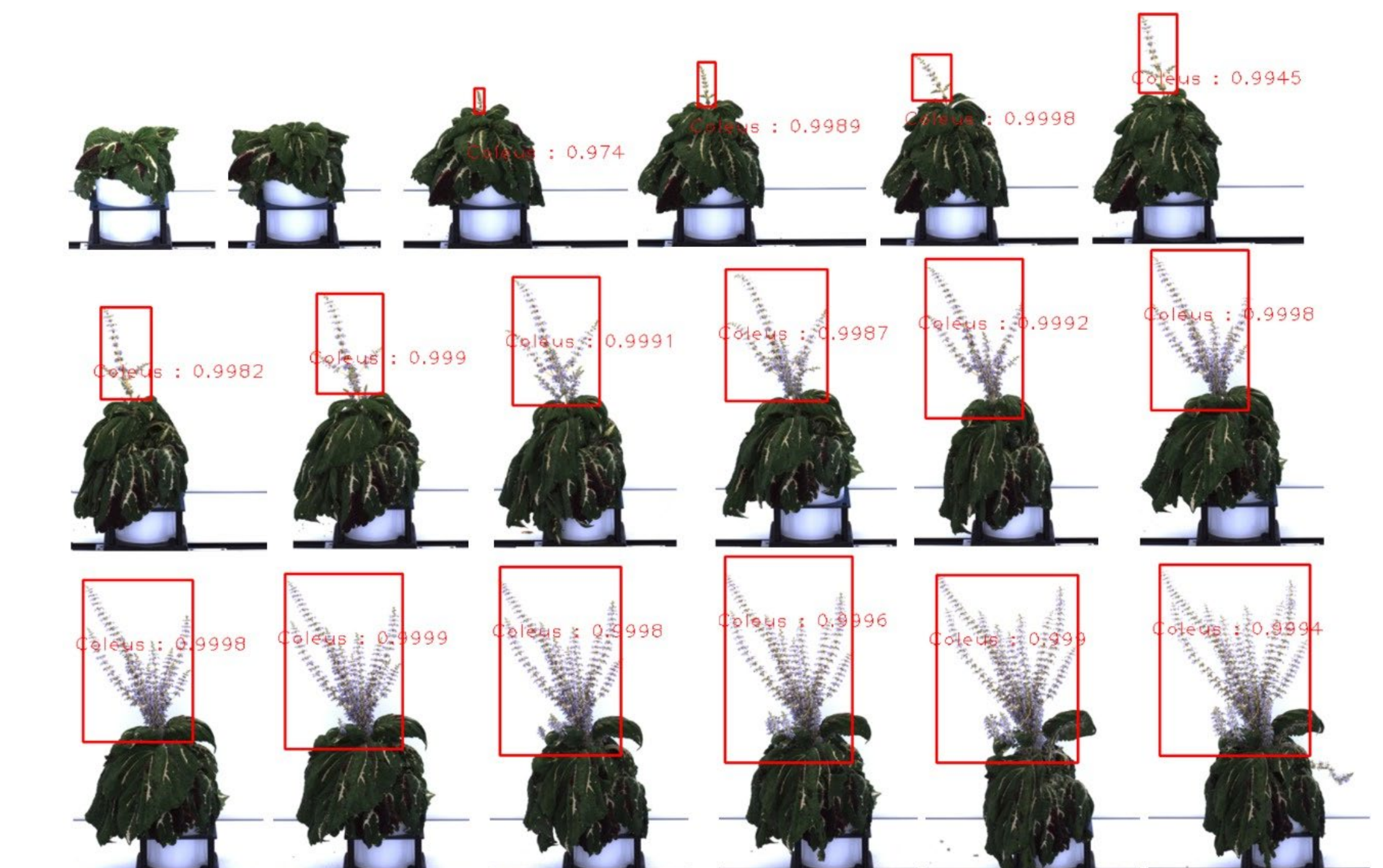
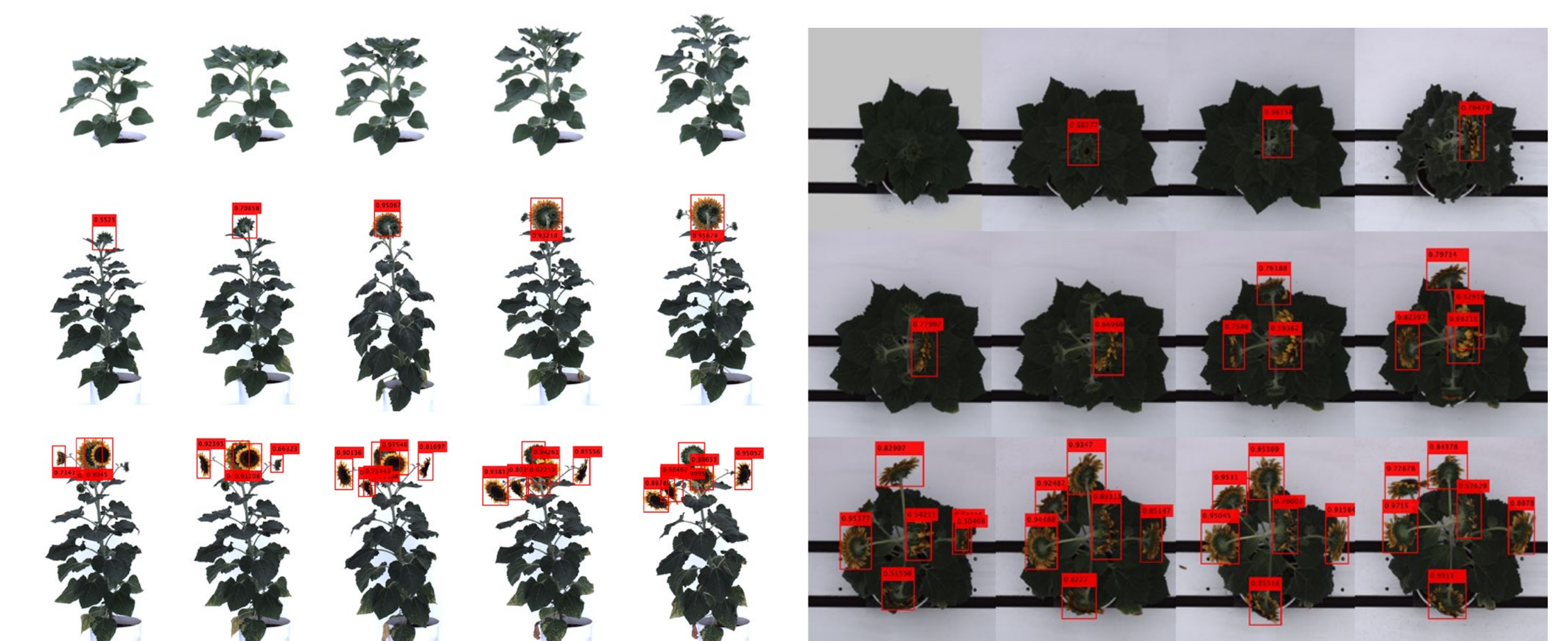


Fig. 7: Illustration of flower emergence timing detection using sample sunflower and coleus image sequences

## FlowerPheno Dataset

- ❖ Consists of RGB image sequences
- ❖ Three flowering plant species: sunflower (*Helianthus annuus*), canna (*Canna generalis*) and coleus (*Plectranthus scutellarioides*)
- ❖ Total number of plants for each category: 20
- ❖ Number of days: 24 to 35
- ❖ Number of views: 10
- ❖ Download from <https://plantvision.unl.edu/dataset>



Fig 8: Sample images from the dataset