Plant Stress Detection Based on Spectral Imaging and Deep Learning

Xiaosheng Luo

Imperial College London

Supervisor: Oliver Windram, Department of Life Sciences (Silwood Park), Imperial College London

Introduction

- 1 Organic growers face a number of challenges they must overcome to deliver high quality food to their
- 2 customers. The lack of reactive chemical control measures in organic farming requires additional levels
- 3 of insight in order to optimise plant and hervest schedule. Traditional manual detection can only wait
- 4 until late crop stress morphological changes, and is generally time-consuming and subjective. However,
- 5 using spectral imaging technology, the differences in crop physiological structure characteristics will
- 6 lead to differences in light reflection, absorption and transmission. Studying these differences in
- 7 spectral imaging can help identify the crop growth status, nutrient and disease.
- 8 Previous studies have made extensive fundamental research on crop detection using spectral technol-
- 9 ogy, mainly based on spectral index features and image features. For instance, the olive Verticillium
- 10 wilt can be diagnosed by images collected from UAV-mounted multi-spectral camera and thermal
- infrared camera, and it was found that early Verticillium wilt was related to green light band, and
- 12 chlorophyll fluorescence index decreased with the disease aggravating (Calderón et al., 2013). More-
- over, applied hyperspectral imaging techniques to identify the angular leaf spot of cucumber by de-
- 14 tecting the contents of chlorophyll and carotenoids showed the feasible for visualizing the pigment
- distribution in cucumber leaves in response to angular leaf spot (Zhao et al., 2016). Besides, progress
- has also been made in the research of crop identification (Torres-Sánchez et al., 2013, Laliberte and
- Rango, 2009), as well as crop growth status monitoring (Vega et al., 2015).
- 18 Although the identification of different plant stresses has reached a considerable level of accuracy, the
- 19 pratical plant stress process may be more complicated and the understanding of the spectral images
- of plant stress could be deeper. In order to explore the feasibility of using spectral imaging to identify
- 21 complex plant stress and better apply spectroscopy technology to pratical production, we propose the
- 22 application and research and as follow:
- 1) How can we build a robust classifier to detecte the quality of organically grown broccoli on
- conveyor belts using spetral imaging?
- 25 2) Can we build a robust classifier based on limited data set, using the spectral images collected
- under different combinations of stress (temperature, light, drought)? How can we extend to the
- 27 field?

28

3) Deep into a metabolic levels, can we find the evidence to support the classification?

keywords: Organic farming, Deep learning, Plant stress, Spectral imaging, Detection, Metabolomics

Methods

- 29 First, use multi-spectral imaging technology combined with deep learning algorithms to detecte the
- 30 quality of organically grown broccoli on conveyor belts.
- 31 Second, build the same detection model using the data collected by the drone.
- Third, under the environment controlled by the laboratory, collect spectral image data under different
- combinations of stress (temperature, light, drought) and explore the possibility of establishing a robust
- 34 classifier in a limited data set.
- Fourth, sample and explore molecular differences by GC-MS, to provide classification basis in metabolic
- 36 levels.

Expected Outcomes

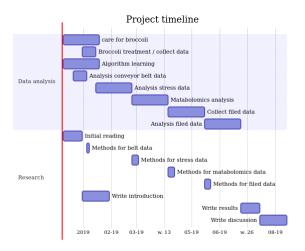
Robust spectral images classifiers and matabolomics analysis.

Budget

£500 funded by school for computing hours, printing costs and for travel related to the project.

Project Feasibility

39 Feasible equipment for data collecting and timeline as bellow



References

- Calderón, R., Navas-Cortés, J. A., Lucena, C. and Zarco-Tejada, P. J. (2013), 'High-resolution airborne hyperspectral and thermal imagery for early detection of verticillium wilt of olive using fluorescence, temperature and narrow-band spectral indices', *Remote Sensing of Environment* 139, 231–245.
- Laliberte, A. S. and Rango, A. (2009), 'Texture and scale in object-based analysis of subdecimeter resolution unmanned aerial vehicle (uav) imagery', *IEEE Transactions on Geoscience and Remote Sensing* 47(3), 761–770.
- Torres-Sánchez, J., Peña-Barragán, J., Gómez-Candón, D., De Castro, A. and López-Granados, F. (2013), Imagery from unmanned aerial vehicles for early site specific weed management, *in* 'Precision agriculture13', Springer, pp. 193–199.
- Vega, F. A., Ramírez, F. C., Saiz, M. P. and Rosúa, F. O. (2015), 'Multi-temporal imaging using an unmanned aerial vehicle for monitoring a sunflower crop', *Biosystems Engineering* **132**, 19–27.
- Zhao, Y.-R., Li, X., Yu, K.-Q., Cheng, F. and He, Y. (2016), 'Hyperspectral imaging for determining pigment contents in cucumber leaves in response to angular leaf spot disease', *Scientific reports* **6**, 27790.