

Agricultural Crop detection, Localisation, and Pose Estimation

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Motivation

Problem: The global population is expected to pass 9 billion by 2050 requiring dramatic scaling of agricultural practices that are difficult with current methods.

Solution (or part of): Agricultural robots (AgBots) are expected to fit into a broader picture in which crop breeding, planting, and harvesting practices are refined to allow greater levels of automation.

Key Roadblock: AgBots require advanced sensor data processing to identify target crops at high speeds.

Aim

To apply the latest advances in robotic vision to the problems of detecting, localising, and estimating the pose of target crops, and to implement our approach on a robotic platform for selective harvesting.

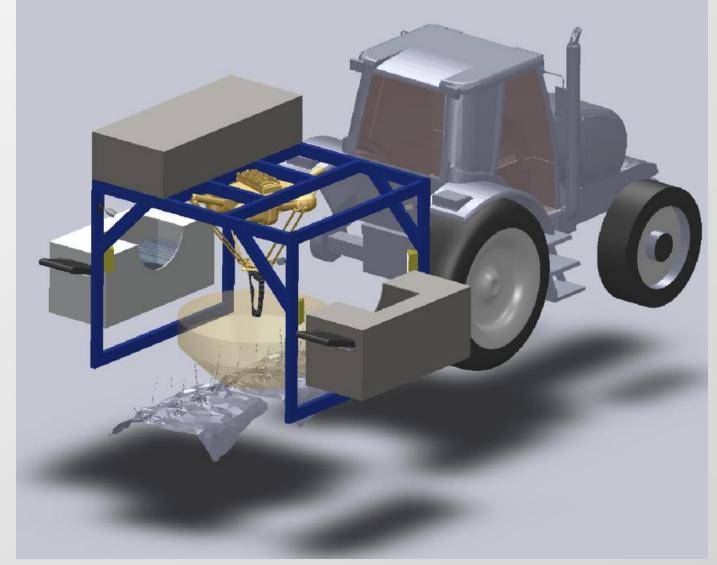
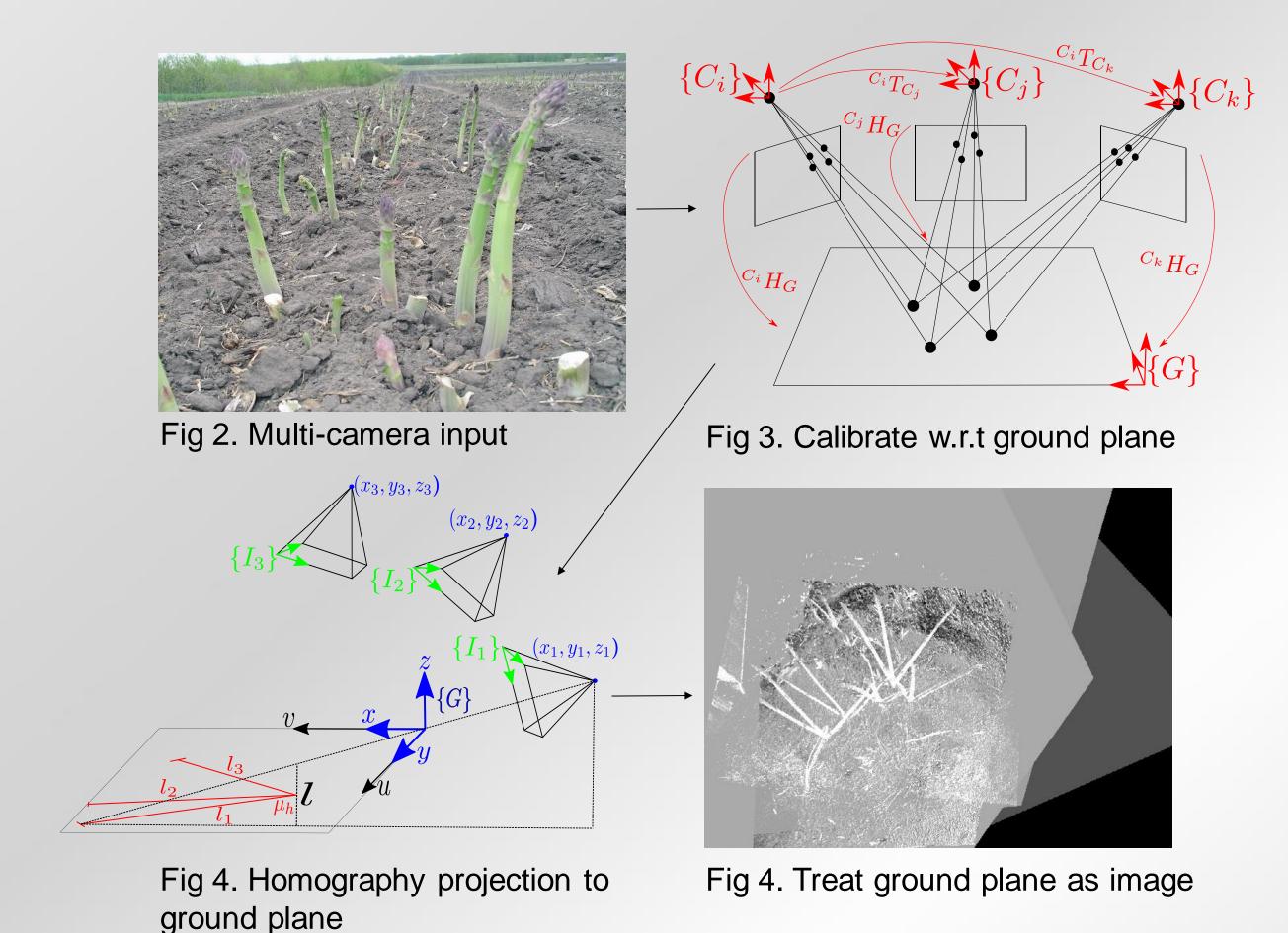


Fig 1. Robotic harvester mounted to tractor

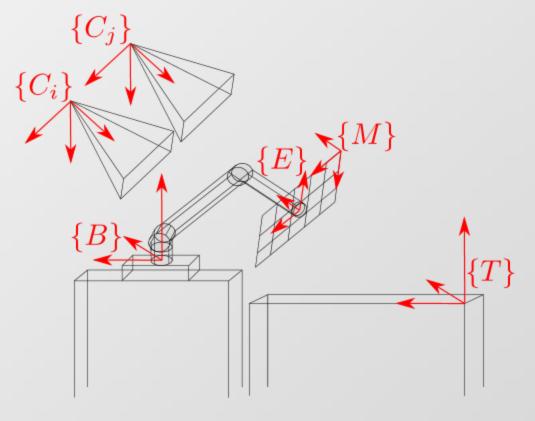
Multi-Camera Data

We have a fixed setup, so let's use it!



Calibration

For successful harvesting we need a multi-camera system calibrated with a robotic manipulator.



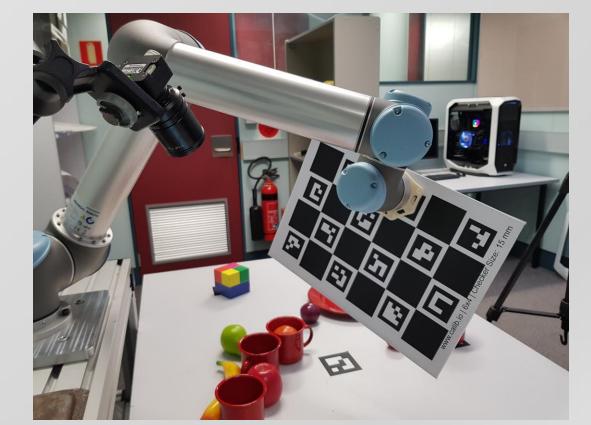
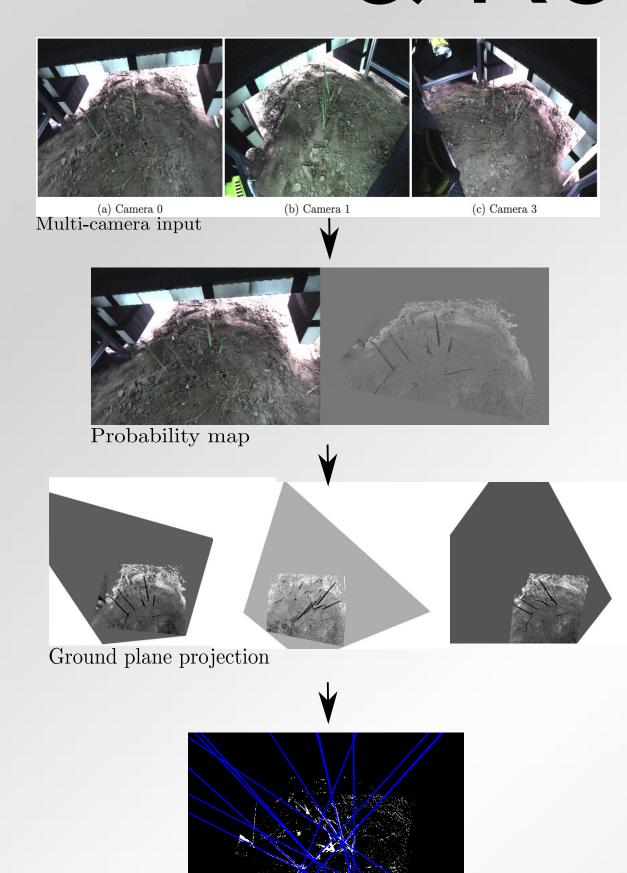


Fig 5. Calibration setup

Fig 6. Calibration of UR5 with multi-cameras

Target poses:
$${}^BX_E(\theta_t)$$
 from forward kinematics ${}^BX_C(\theta_t) = {}^BX_E(\theta_t) \cdot {}^EX_C$ ${}^BX_T = {}^B\bar{X}_C \cdot {}^CX_M \cdot {}^MX_T$

Initial Implementation & Results



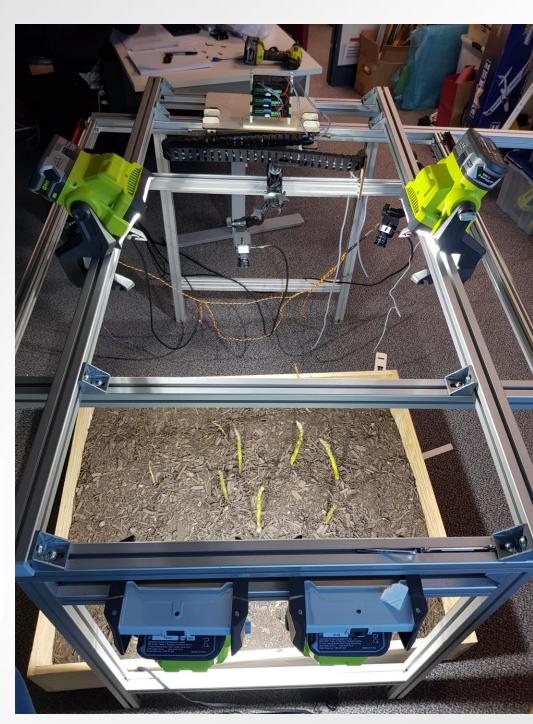


Fig 8. Rig for in-lab testing

Experiment	RMSE (mm)
1	0.137
2	1.829
3	0.128
4	1.778
5	0.207
6	0.333
Overall	0.735

Fig 7. Initial Implementation

Intersection Detection & Tracking

Fig 9. Initial in-lab results

Future Work

Short term: Do ground-plane images from multiple views (Fig. 4) provide a useful input to learn crop pose?

Long term: Robustify perception pipeline and implement on a robotic platform via a distributed architecture for on-farm harvesting.

