

## 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.

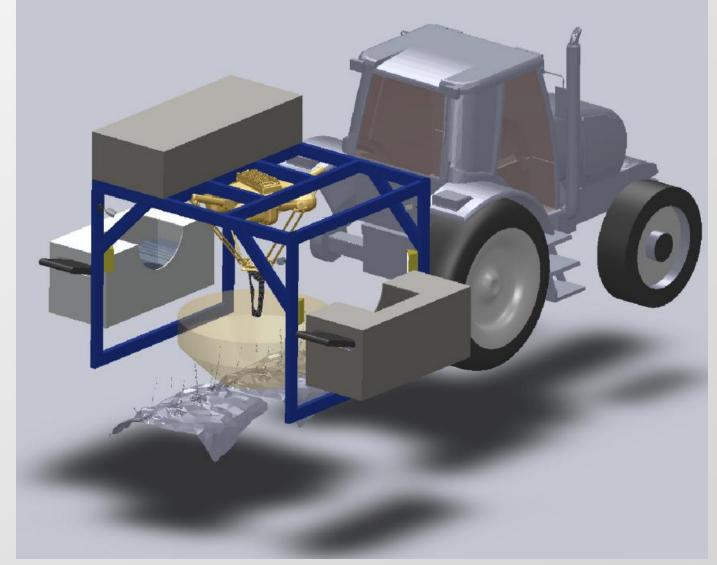
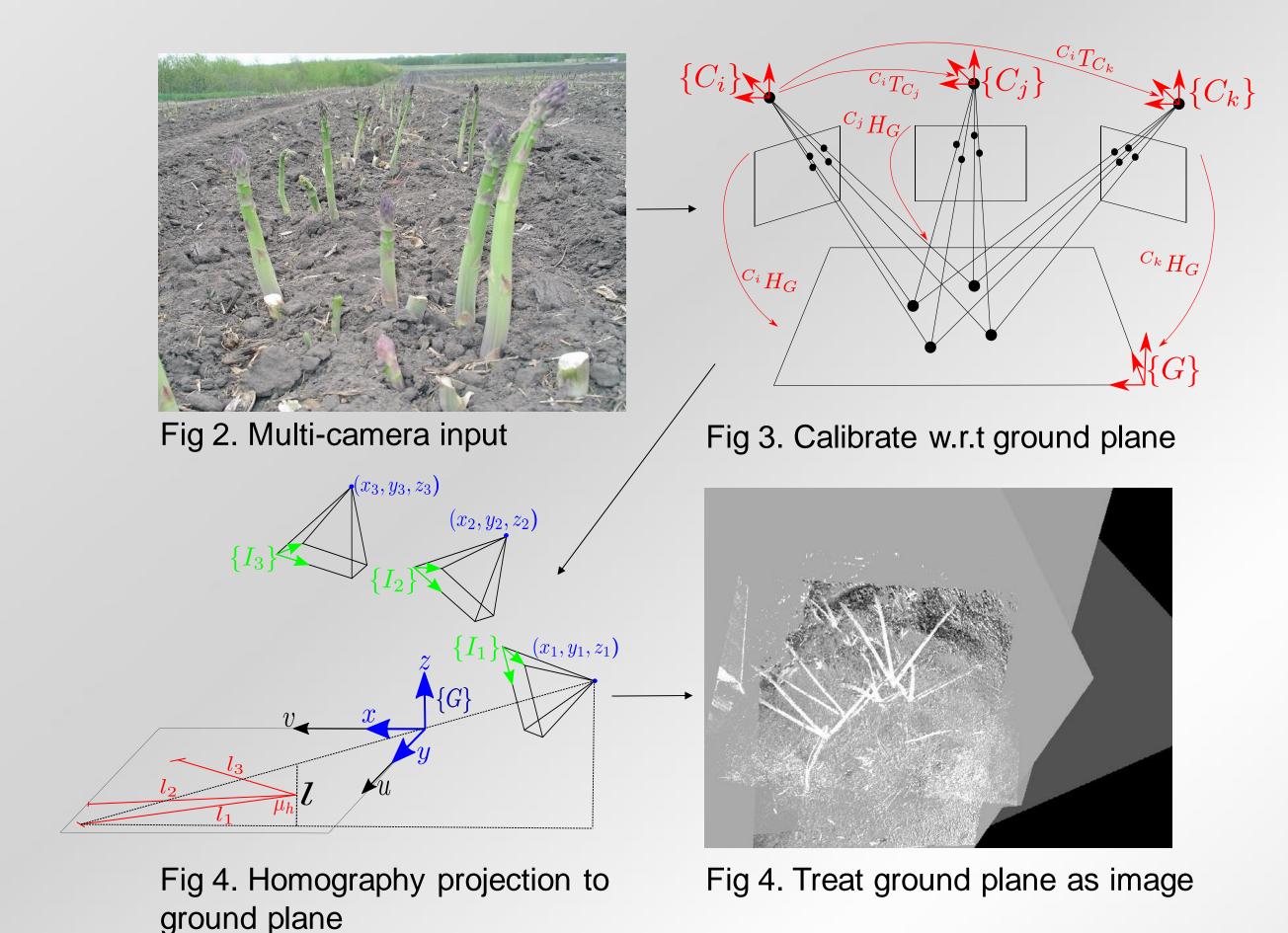


Fig 1. Robotic harvester mounted to tractor

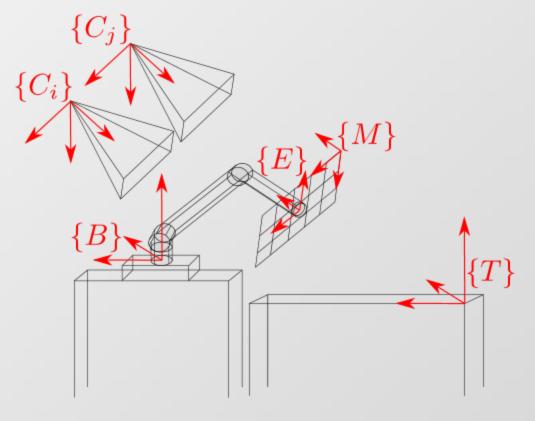
### Multi-Camera Data

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.



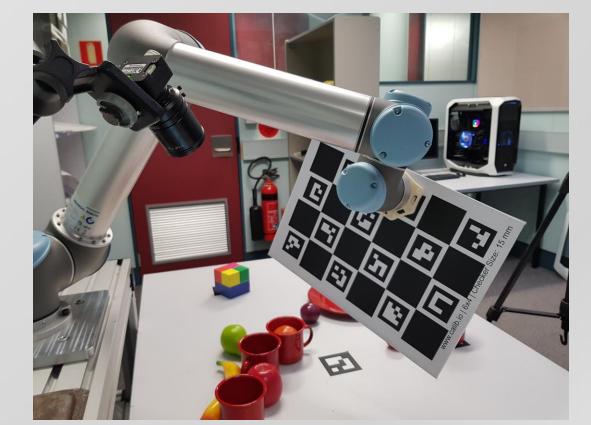
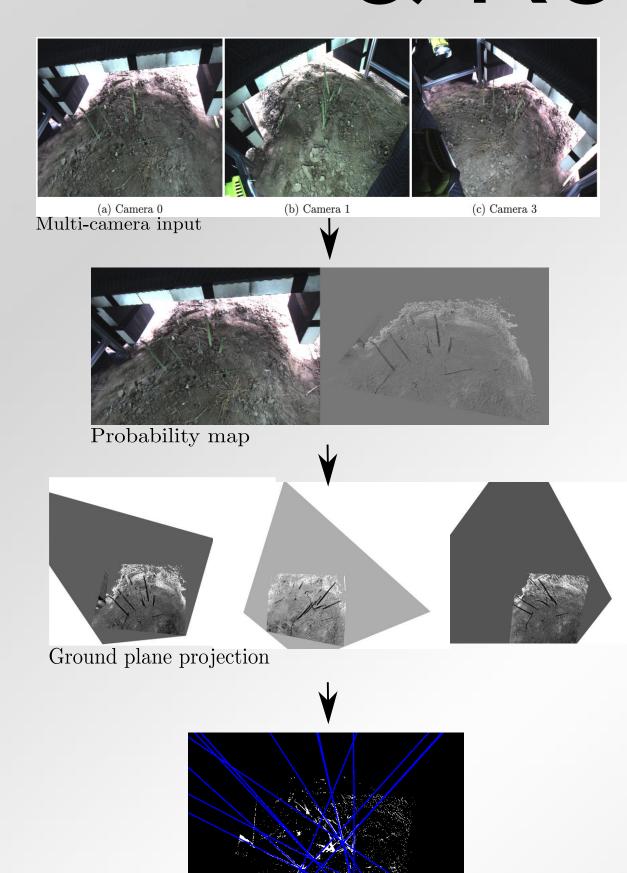


Fig 5. Calibration setup

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



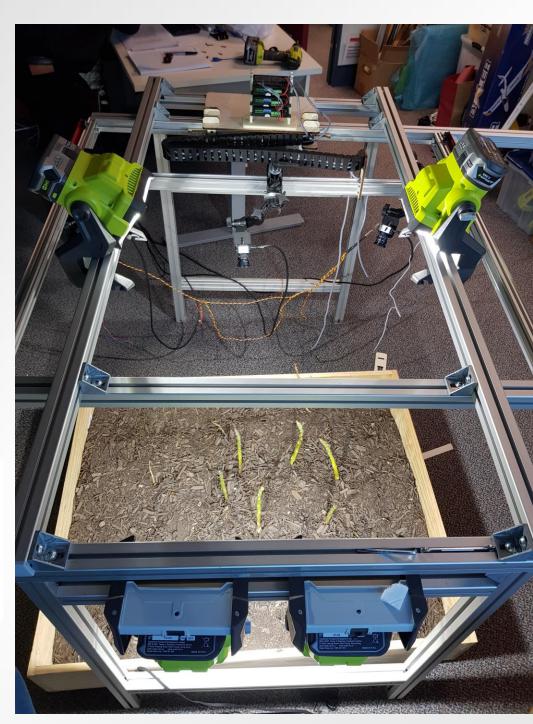


Fig 8. Rig for in-lab testing

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.

