

Introduction – Robotic Green Fruit Thinning

- Green fruit is one of the most laborintensive tasks of apple production.
- A robotic thinning system could reduce the high labor requirement.
- The end-effector and vision system are important components for robotic thinning.
- Green fruit removal dynamics and an end-effector prototype were investigated.
- Green fruit segmentation and orientation estimation algorithms are currently being investigated.







- The dynamics of green fruit removal using pulling and stem cutting methods were investigated.
- Stem cutting required significantly more force than pulling, although pulling resulted in a high occurrence of spur-end stem detachment.
- No correlation between fruit and stem dimensions and required removal force was found.

End-Effector – Stem-Cutting Prototype





- A stem-cutting end-effector prototype was evaluated in thinning experiments using a handheld bar and robotic manipulator.
- ♣ Fruit removal success rate was $\geq 90\%$ for all experiments.

Development of Robotic Green Fruit Thinning System for Apple Orchards

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Vision System – Orientation Estimation



- fruit and stems.

Conclusions

- thinning system.

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Vision System - Segmentation





Mask R-CNN used for green fruit and stem segmentation.

365 training images, 78 validation images, and 78 test images were used. Test dataset average precision scores (IoU = 0.5) for green fruit and stems are 83.5% and 38.7%, respectively.

Object sizes in images can affect performance.





Principal Component Analysis algorithm used to estimate the orientation of green

Mean absolute errors for green fruit and stem orientations over 57 images are 15.2° and 3.4°, respectively.

Occlusions reduce the accuracy of orientation estimations.

The stem-cutting end-effector prototype is a simple yet effective device that could be used by a robotic green fruit thinning system.

The green fruit segmentation and orientation estimation algorithms could serve as bases for further development of a vision system for robotic green fruit thinning.

Future work will be done in path planning for robotic manipulator in robotic fruit





