



Towards an AI-based device for automatic in-field fruit counting and sizing

Jordi Gené-Mola, Marc Felip-Pomés, Francesc Net-Barnés, Josep-Ramon Morros, Juan C. Miranda, Jaume Arnó, Luís Asín, Jaume Lordan, Javier Ruiz-Hidalgo, Eduard Gregorio

Efficient Use of Water in Agriculture Program, Institute of AgriFood, Research and Technology (IRTA), Lleida, Catalonia, Spain

Research Group in AgrolCT& Precision Agriculture GRAP, Department of Agricultural and Forest Sciences and Engineering, Universitat de Lleida (UdL) Agrotecnio-CERCA Center, Lleida, Catalonia, Spain

Department of Signal Theory and Communications, Universitat Politècnica de Catalunya Barcelona, Catalonia, Spain

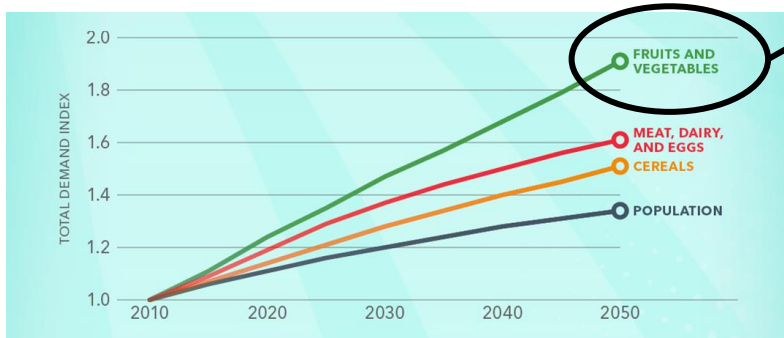
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Introduction



How can we meet it?

- **Optimization** of orchards management
- Need to **monitor the fruit production**: number and size of fruits



AI-based solutions

- High cost



Manual measurements

- Need of sampling (source of error)
- Time-consuming
- Labour-intensive



Our solution



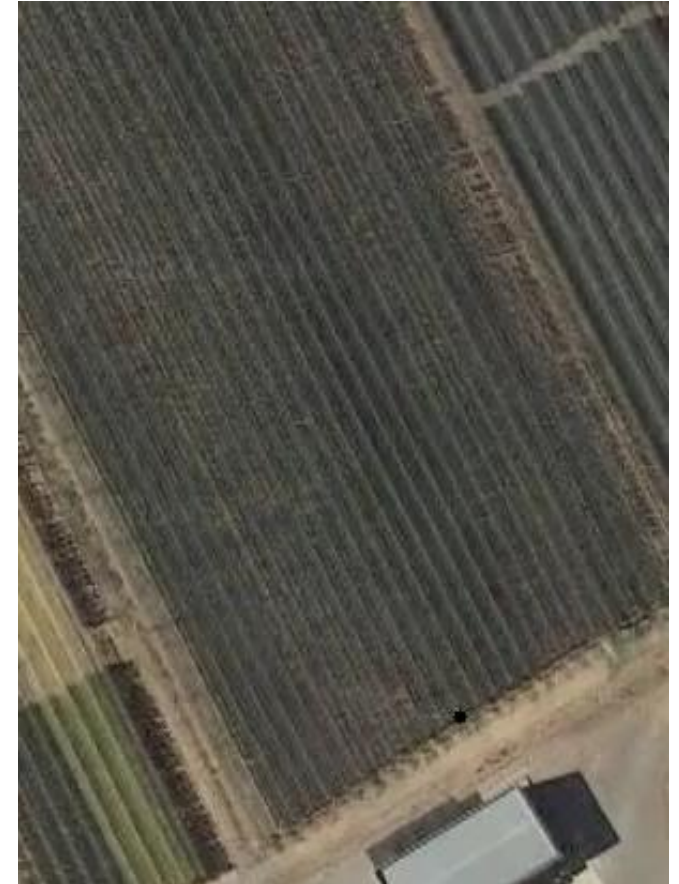
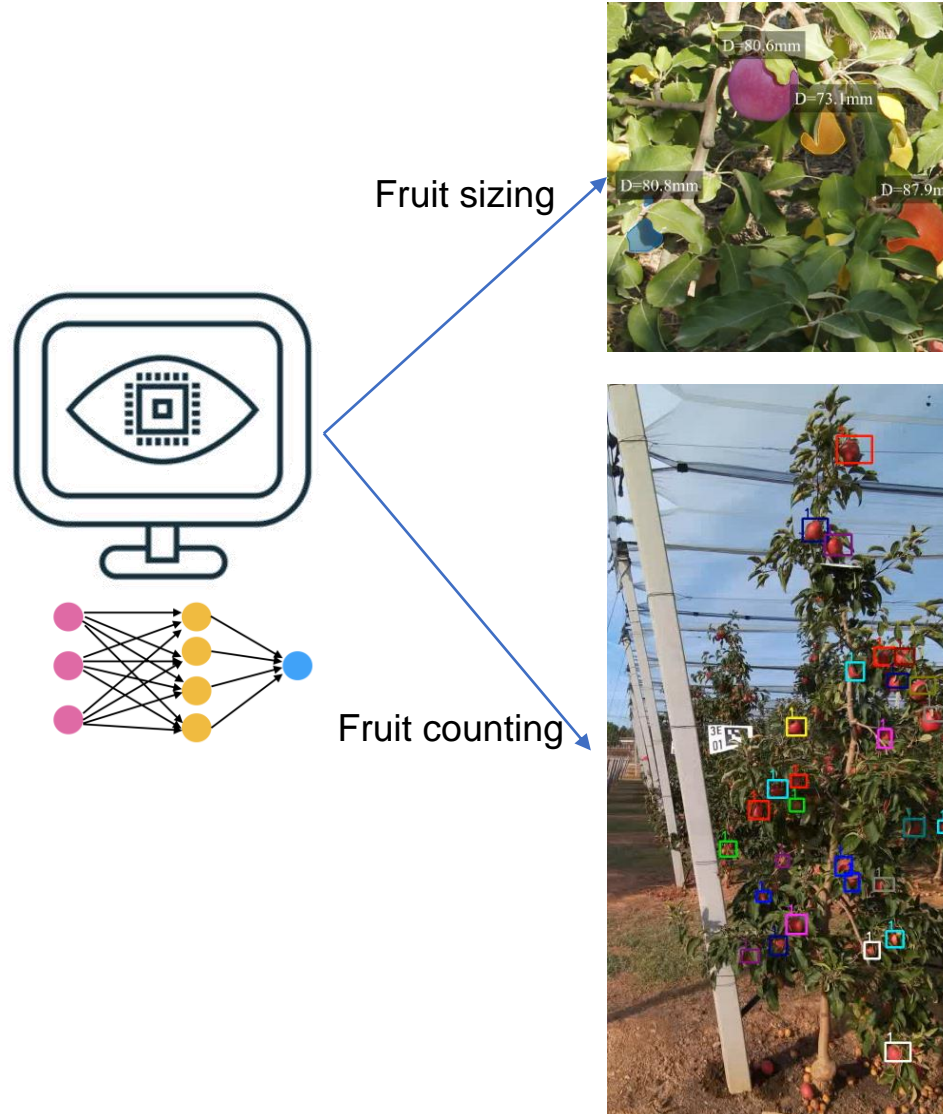
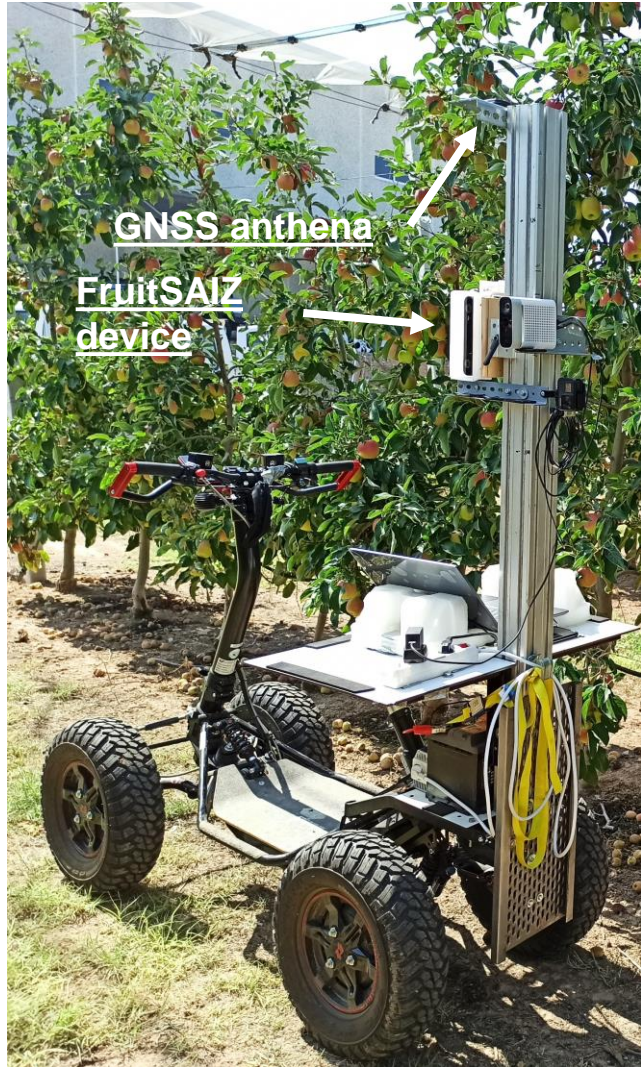
Source: IFPRI (International Food Policy Research Institute). "International Model for Policy Analysis of Agricultural Commodities and Trade (IMPACT)." 2017 Global Food Policy Report (2017): 110-118

Source: <https://apal.org.au/pips-3-green-atlas-cartographer/>

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The FruitsAlz device

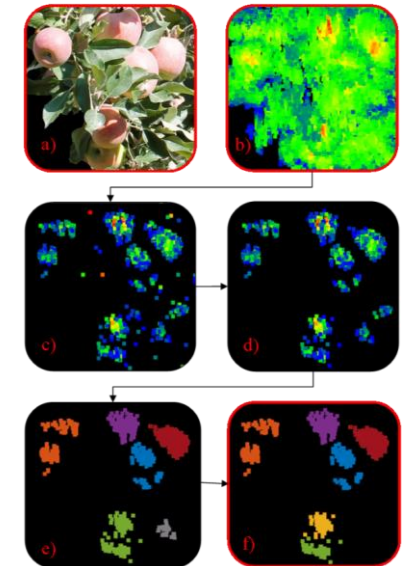
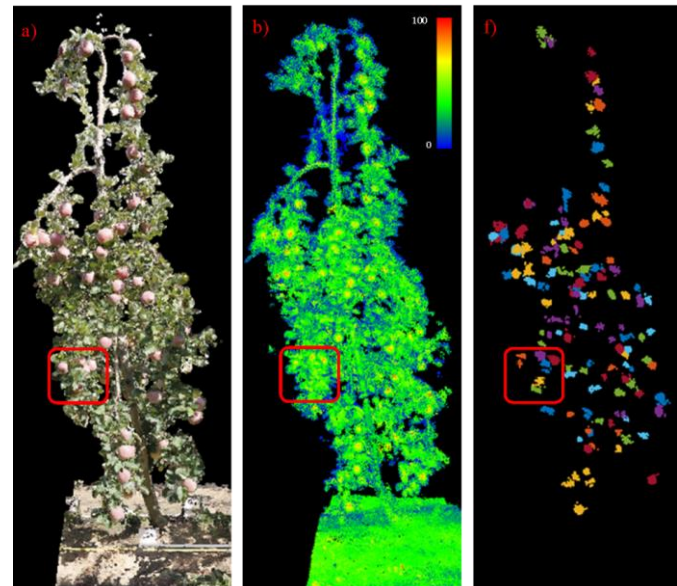


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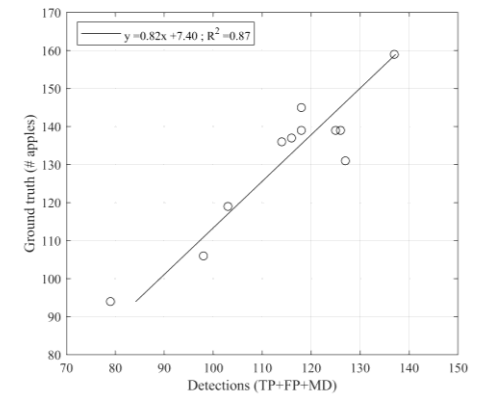
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Background

LiDAR Fruit detection (2019)

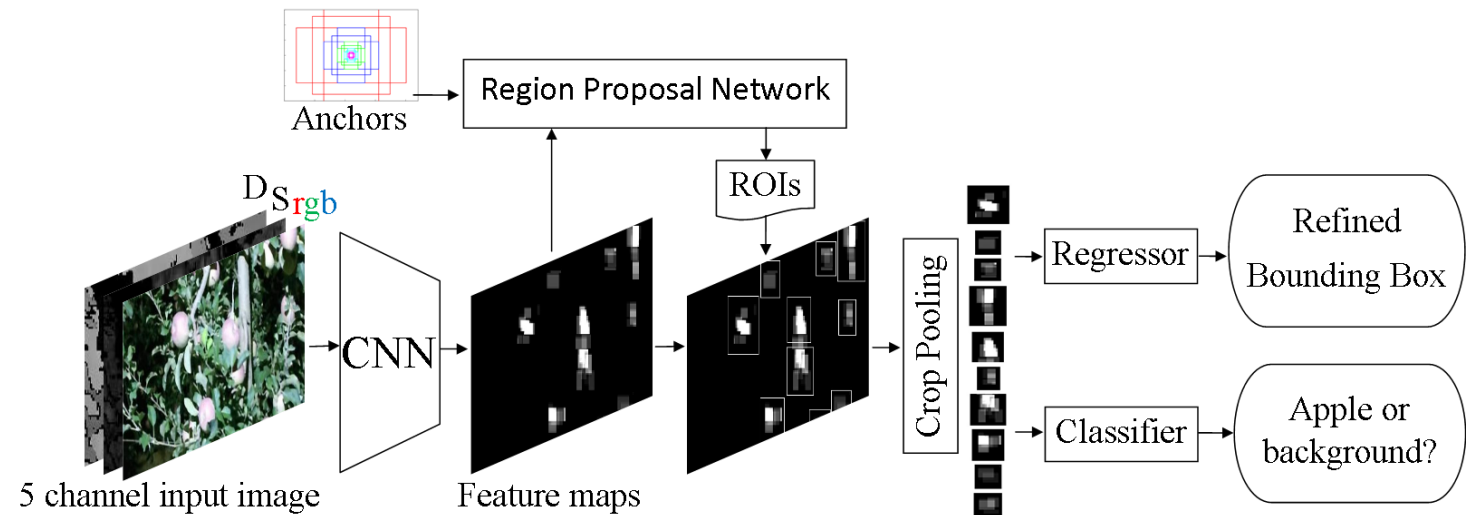
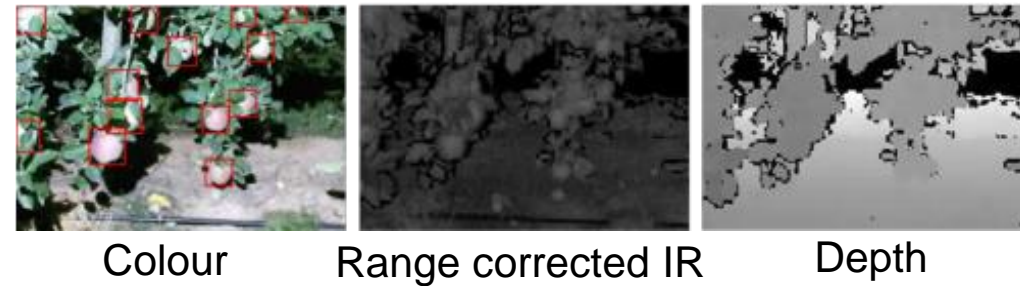


| R | P | $F1\text{-score}$ | R^2 |
|-------|-------|-------------------|-------|
| 0.751 | 0.860 | 0.802 | 0.87 |



Background

Fruit detection using RGB-D cameras (2019)



Fruit counting

Which tree has more fruits?

a)



23 visible apples
130 apples
17.7 % visible

b)



45 visible apples
124 apples
36.3% visible

c)



37 visible apples
119 apples
31.1% visible

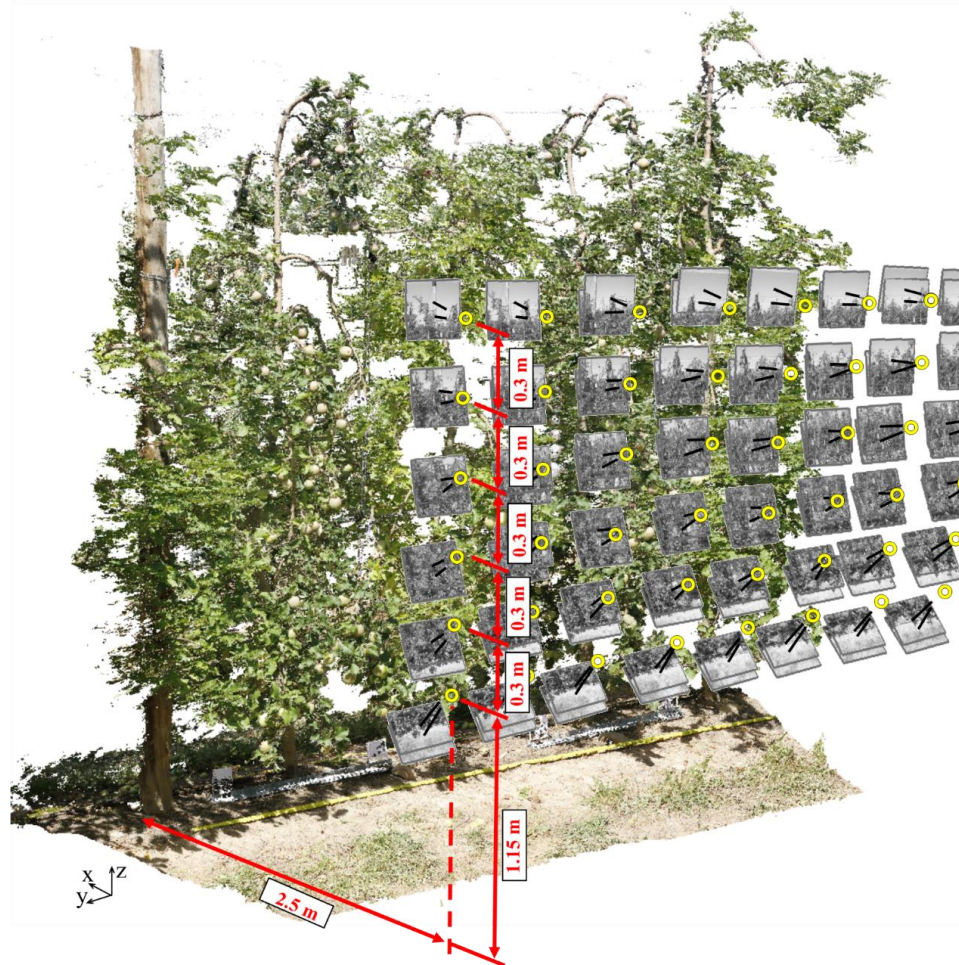
d)



17 visible apples
109 apples
15.6 % visible

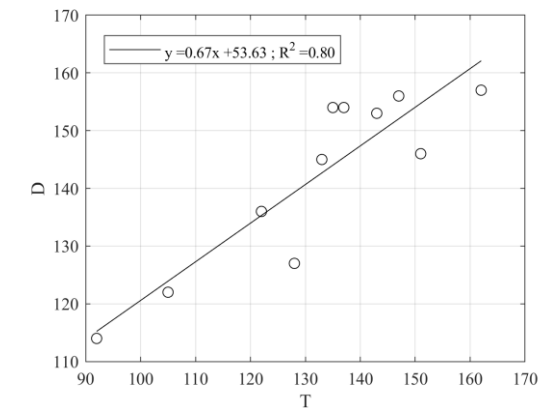
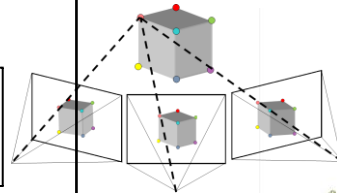
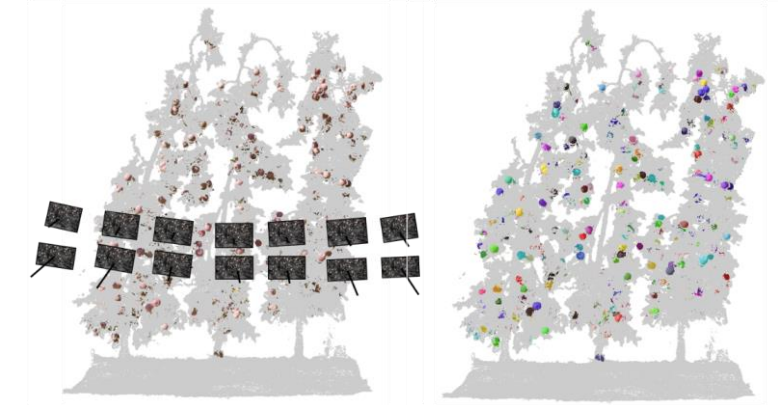
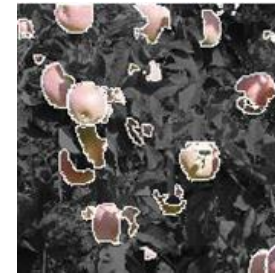
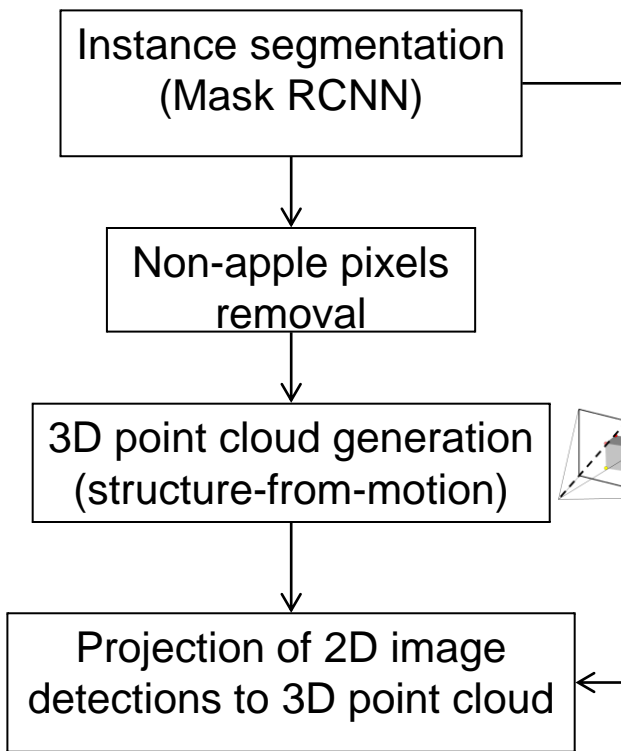
Fruit counting

Hypothesis → Multi-view sensing will increase the number of fruits visible in images



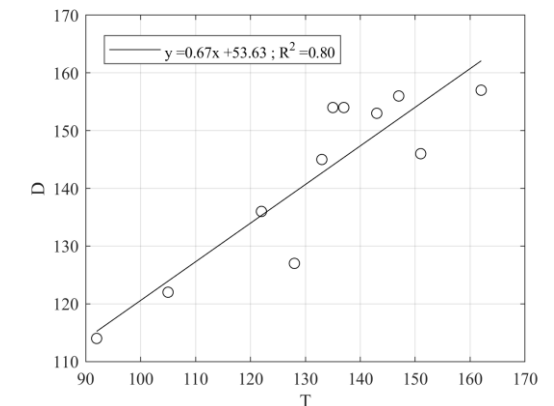
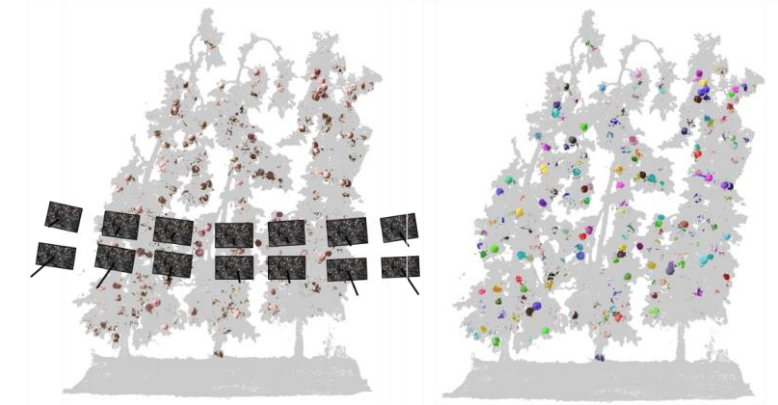
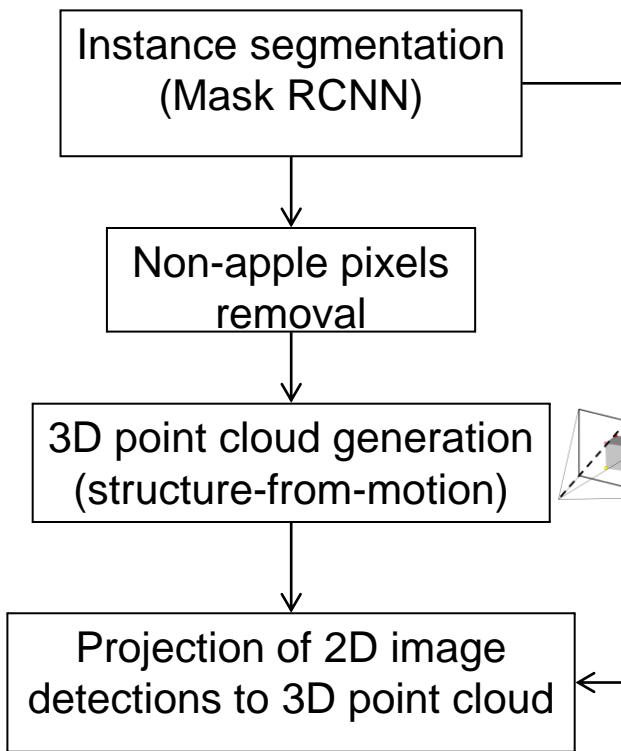
Background

Fruit detection using SfM (2020)

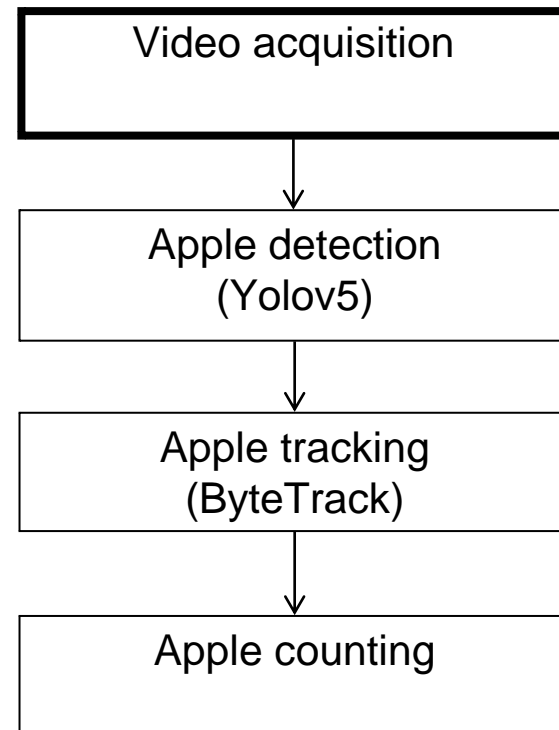


Background

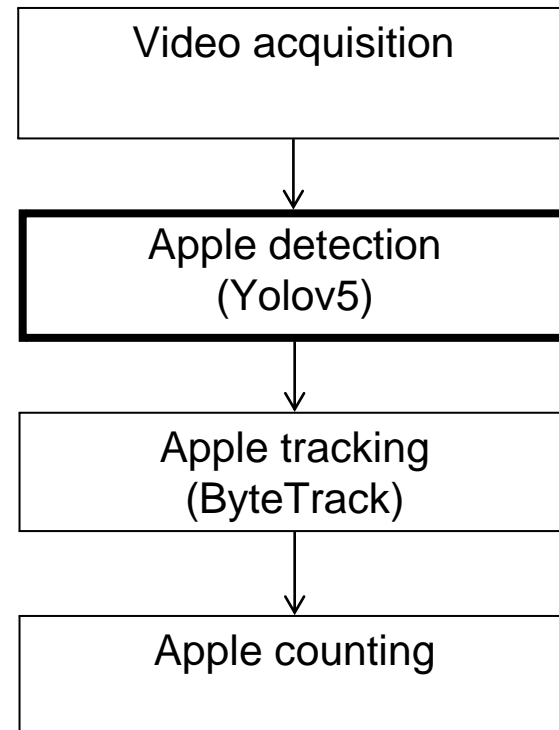
Fruit detection using SfM (2020)



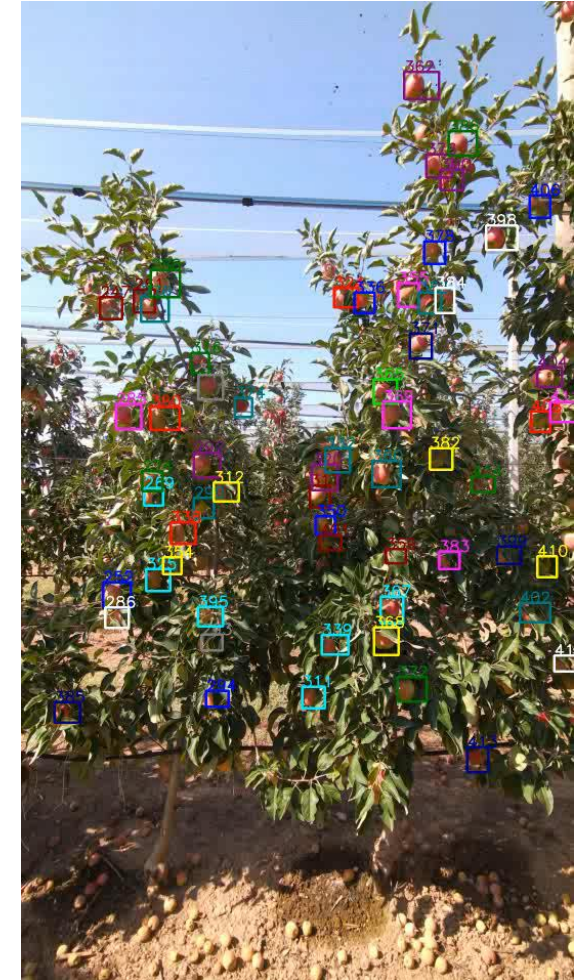
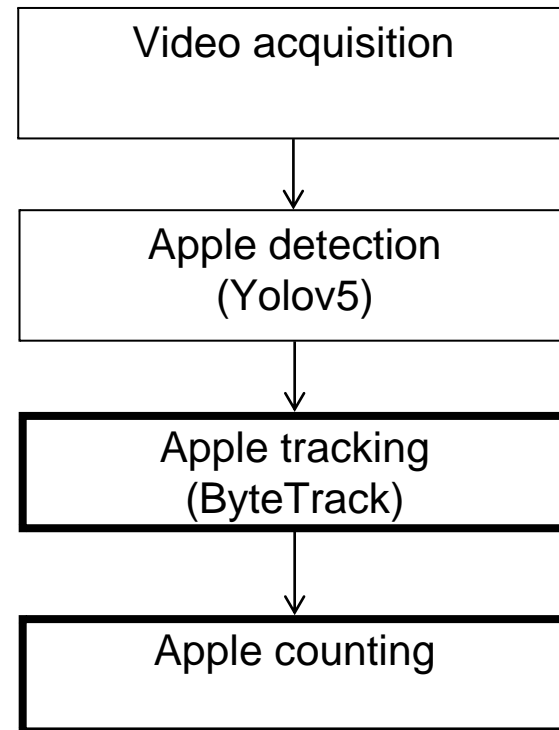
Methodology



Methodology

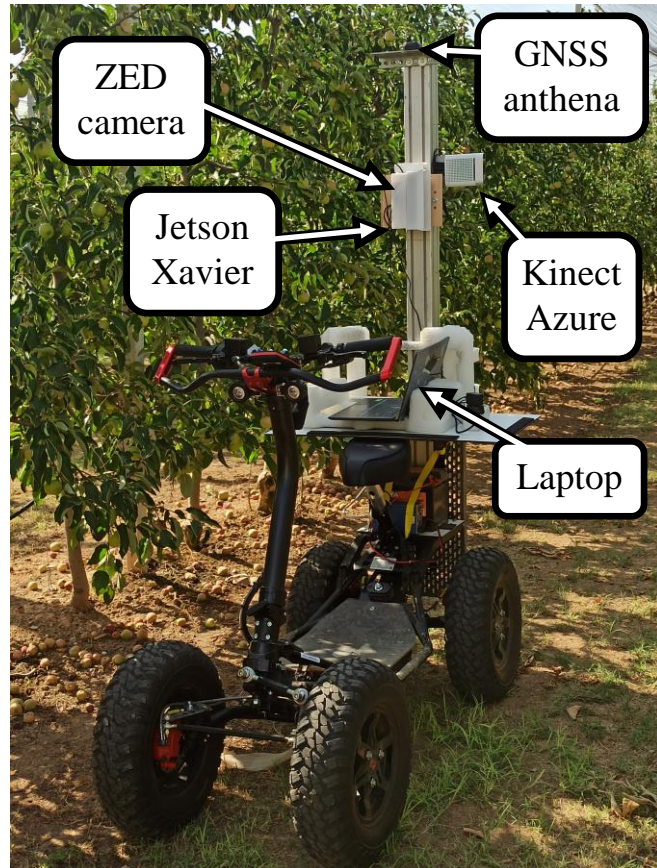


Methodology



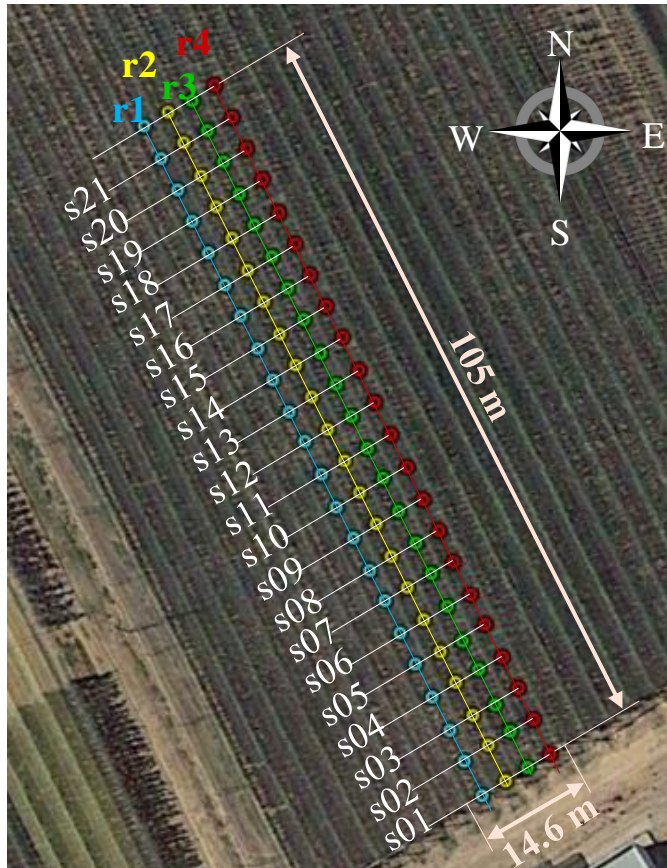
Fruit counting

Experimental set-up



Fruit counting

Experimental set-up. Ground truth



| Fruit load | Mean size (mm) | Mean weight (g) | Yield (kg) |
|-----------------|----------------|-----------------|--------------|
| 603 488 485 611 | 73 72 73 73 | 155 154 157 156 | 93 75 76 95 |
| 204 652 507 497 | 77 69 71 70 | 180 132 145 140 | 37 86 73 69 |
| 452 749 597 482 | 77 72 73 70 | 176 146 154 136 | 79 109 92 66 |
| 522 613 | 77 74 | 176 161 | 92 98 |
| 587 636 404 | 75 75 75 | 168 166 166 | 99 106 67 |
| 524 493 532 | 77 72 72 | 180 155 149 | 94 76 79 |
| 525 467 431 398 | 72 69 73 73 | 148 130 155 155 | 78 61 67 62 |
| 421 618 493 451 | 76 74 74 77 | 174 157 158 179 | 73 97 78 81 |
| 521 601 438 476 | 72 76 72 75 | 150 168 151 162 | 78 101 66 77 |
| 404 512 439 485 | 71 74 76 74 | 149 158 175 154 | 60 81 77 75 |
| 291 435 453 510 | 79 75 74 78 | 193 167 165 177 | 56 73 75 90 |
| 445 461 388 488 | 78 77 77 74 | 183 176 179 152 | 81 81 70 74 |
| 491 465 348 415 | 75 77 81 78 | 174 179 205 189 | 85 83 71 78 |
| 490 436 475 423 | 75 75 76 77 | 167 164 169 177 | 82 72 80 75 |
| 340 510 428 448 | 76 72 73 75 | 174 152 151 166 | 59 78 65 74 |
| 454 416 439 412 | 76 73 72 74 | 168 161 154 162 | 76 67 67 67 |
| 458 386 525 466 | 77 76 73 74 | 177 175 158 161 | 81 68 83 75 |
| 404 428 452 400 | 77 76 72 75 | 177 174 153 167 | 72 74 69 67 |
| 395 417 451 423 | 75 75 73 76 | 166 165 155 175 | 66 69 70 74 |
| 356 388 420 391 | 76 72 69 72 | 170 150 135 150 | 61 58 57 59 |
| 466 512 355 372 | 75 69 74 75 | 168 135 162 166 | 78 69 57 62 |

Fruit counting

Experimental set-up

Kinect Azure



06/09/2021



28/09/2021

Fruit counting

Results

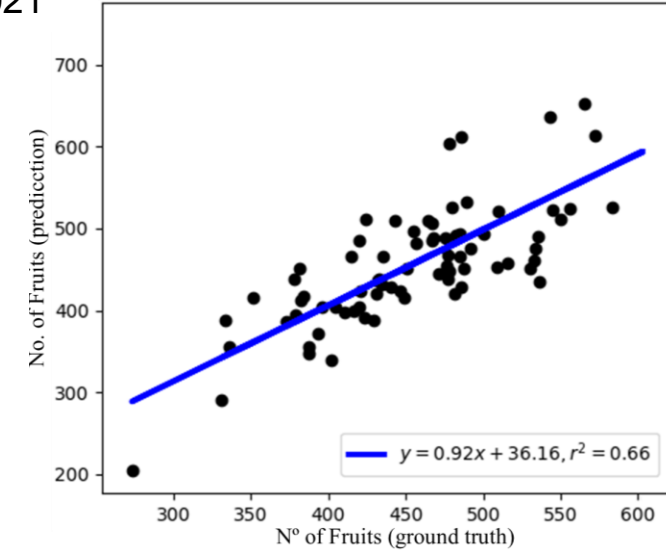
Tracking Evaluation

| <i>Tracker</i> | <i>MOTA</i> | <i>IDF1</i> | <i>HOTA</i> | <i>Time [ms/img]</i> |
|----------------|---------------|---------------|---------------|----------------------|
| SORT | 0.6401 | 0.8091 | 0.6804 | 15.3 |
| DeepSORT | 0.5735 | 0.7646 | 0.6824 | 128.0 |
| ByteTrack | 0.6817 | 0.8369 | 0.6894 | 15.4 |

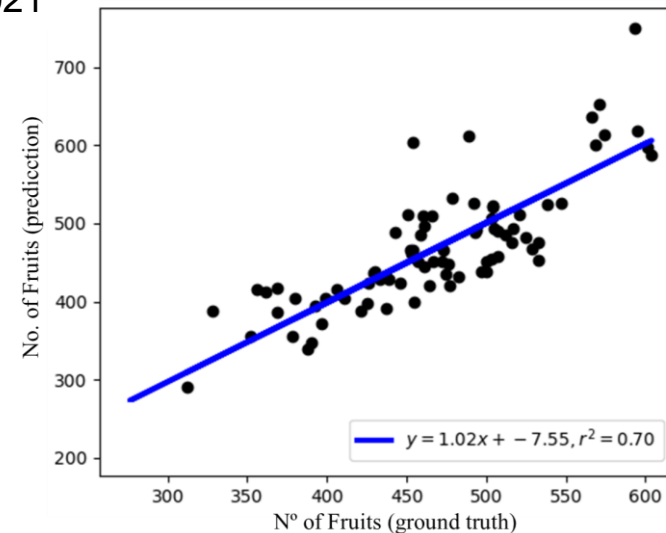
Fruit counting evaluation

| <i>Date</i> | <i>MAE (apples)</i> | <i>MBE (apples)</i> | <i>MAPE (%)</i> | <i>RMSE (apples)</i> | <i>R²</i> |
|-------------|---------------------|---------------------|-----------------|----------------------|----------------------|
| 06/09/2021 | 38.43 | -0.20 | 8.45 % | 49.04 | 0.66 |
| 28/09/2021 | 34.98 | 0.15 | 7.47% | 45.64 | 0.70 |

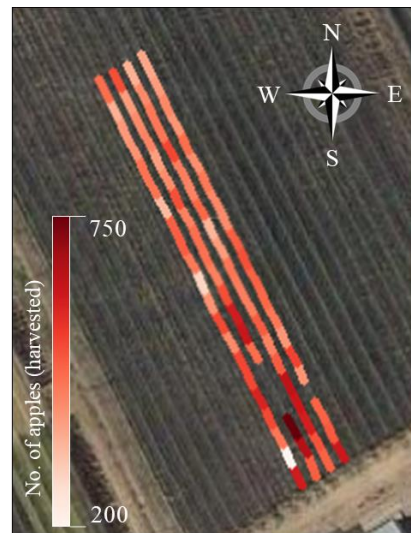
06/09/2021



28/09/2021



Results



Ground Truth

06/09/21

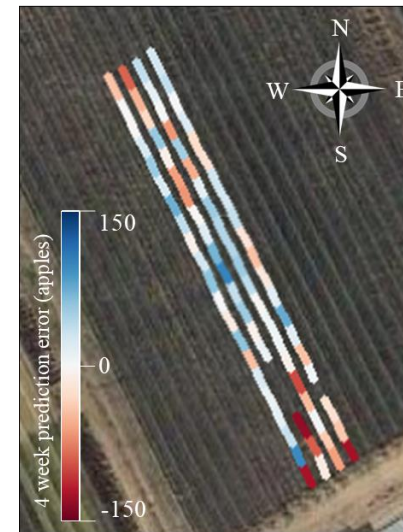


28/09/21

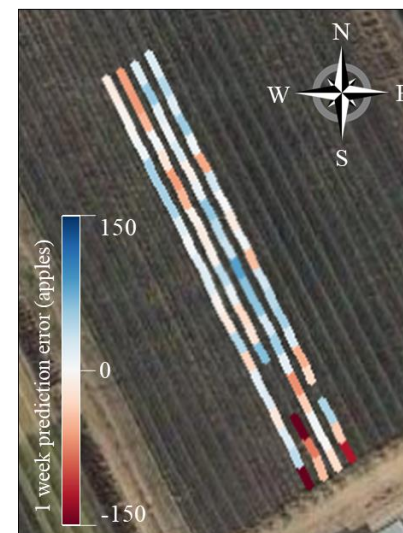


Prediction

06/09/21



28/09/21



Error

Fruit counting

Results

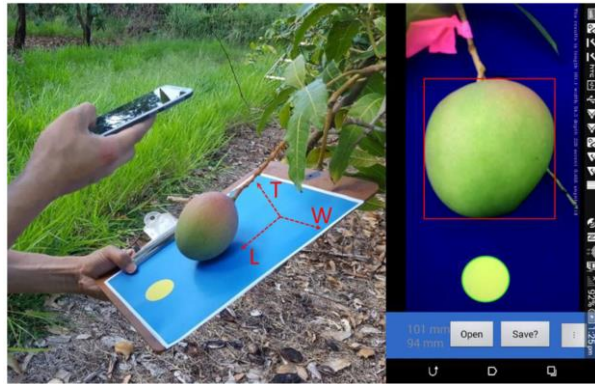


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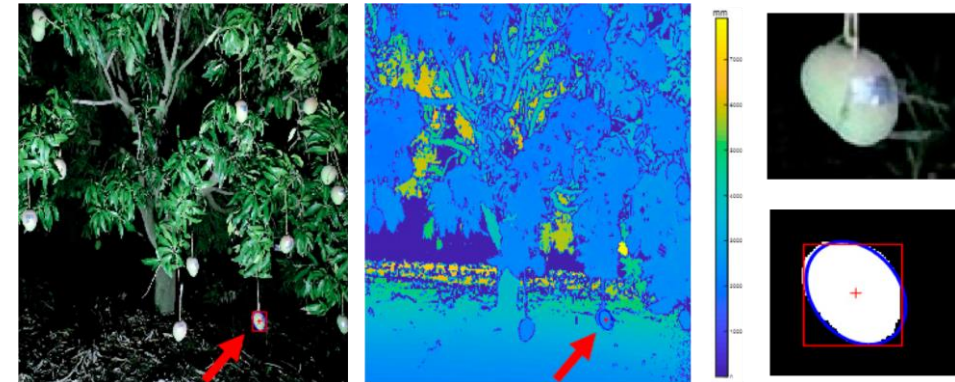
Background

RGB



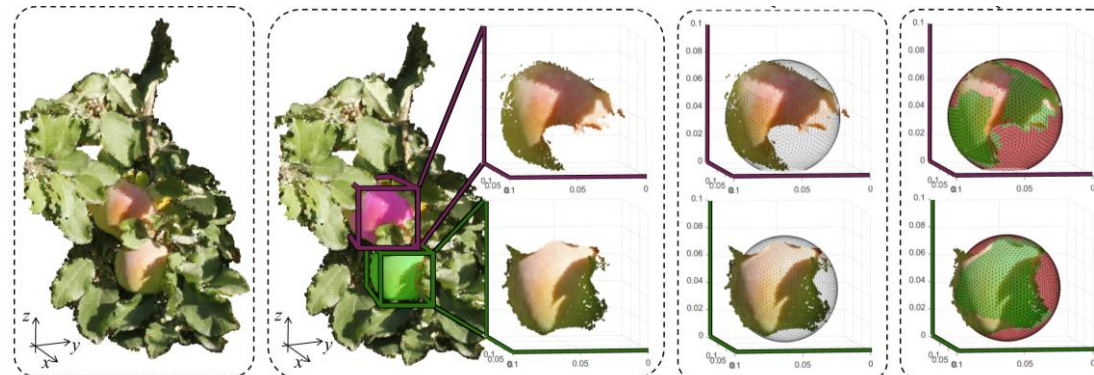
[Wang et al. \(2018\). https://doi.org/10.3390/s18103331](https://doi.org/10.3390/s18103331)

RGB-D



[Wang et al. \(2017\). https://doi.org/10.3390/s17122738](https://doi.org/10.3390/s17122738)

3D

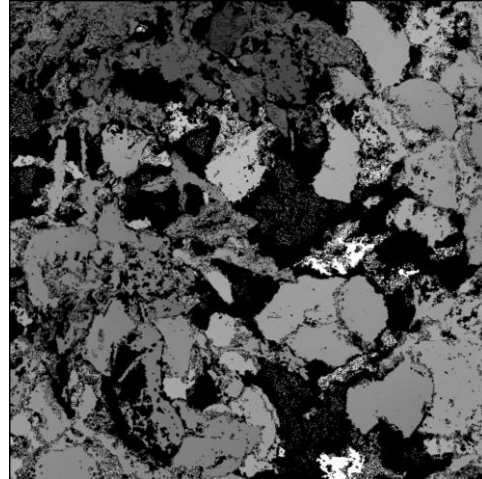


[Gené-Mola et al. \(2021\). https://doi.org/10.1016/j.compag.2021.106343](https://doi.org/10.1016/j.compag.2021.106343)

Background



RGB



Depth



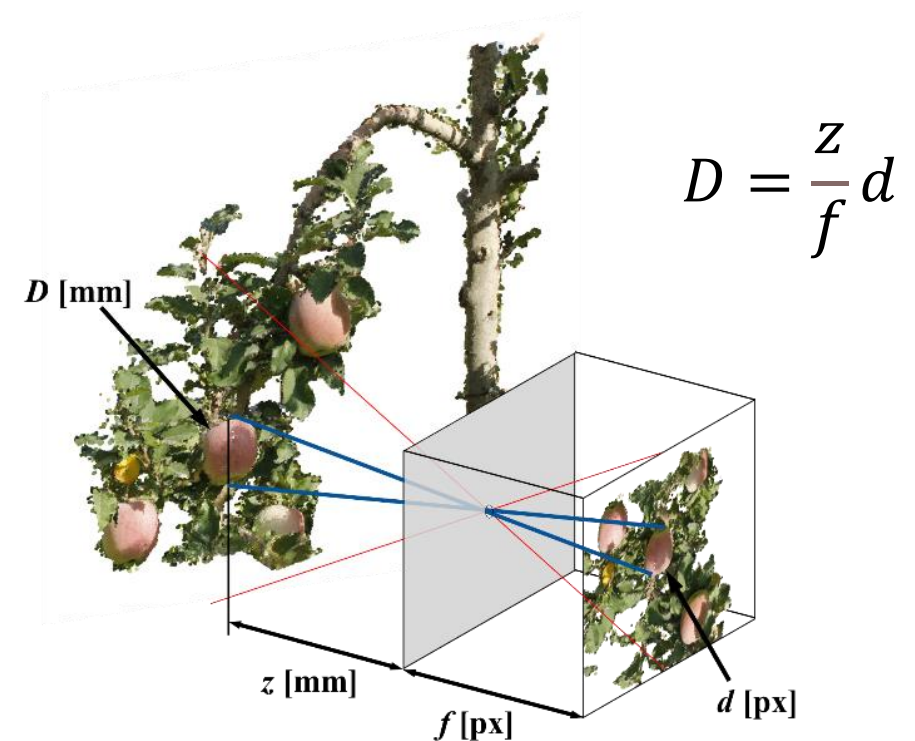
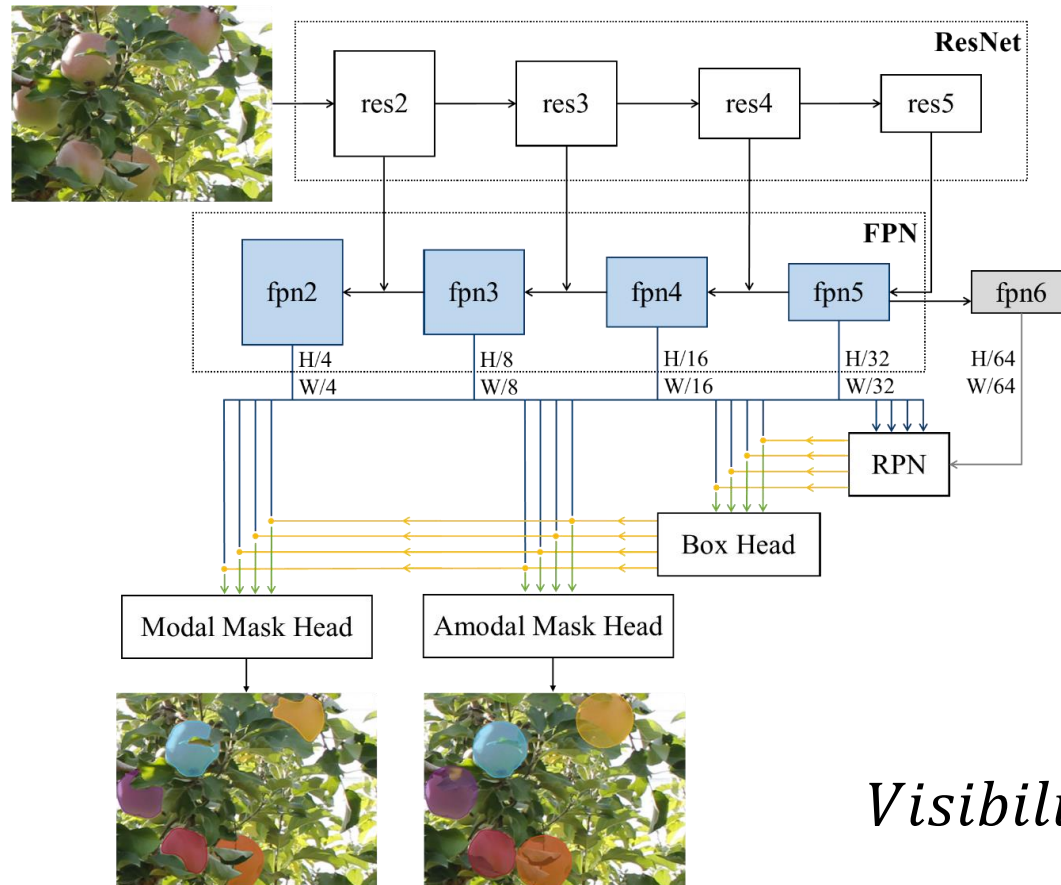
Modal



Amodal

Methodology

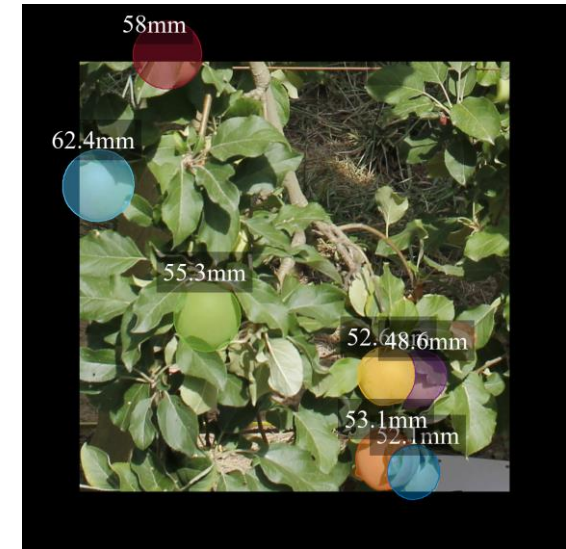
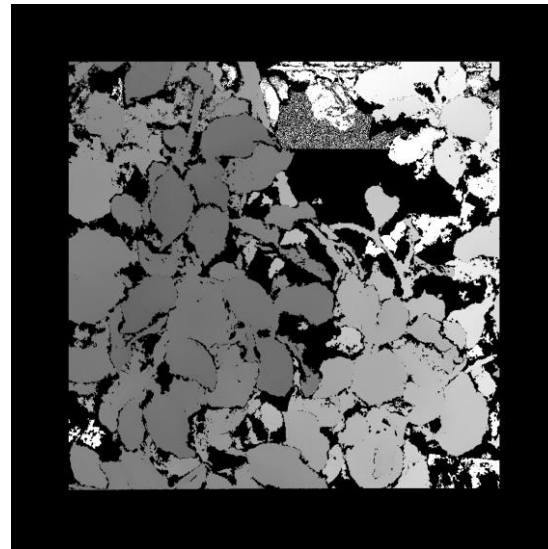
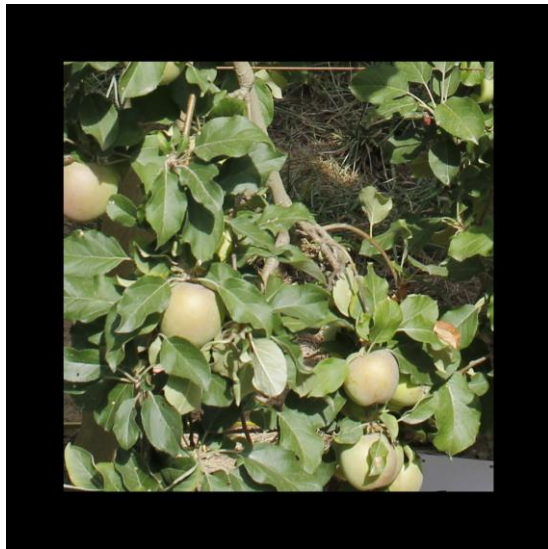
Modified Mask R-CNN



$$Visibility = \frac{A_{modal}}{A_{amodal}}$$

Experimental set-up

Papple RGB-D-Size dataset (Used for Training, Validation and Test)



https://www.grap.udl.cat/en/publications/papple_rgb-d-size-dataset/

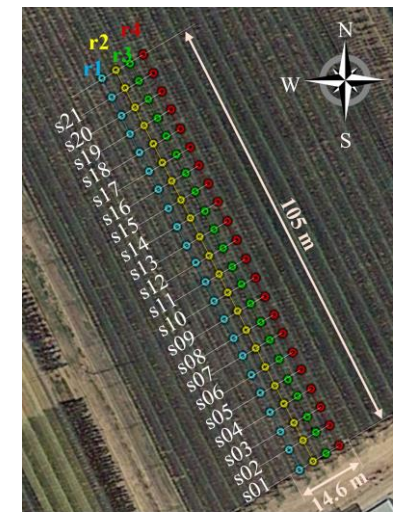
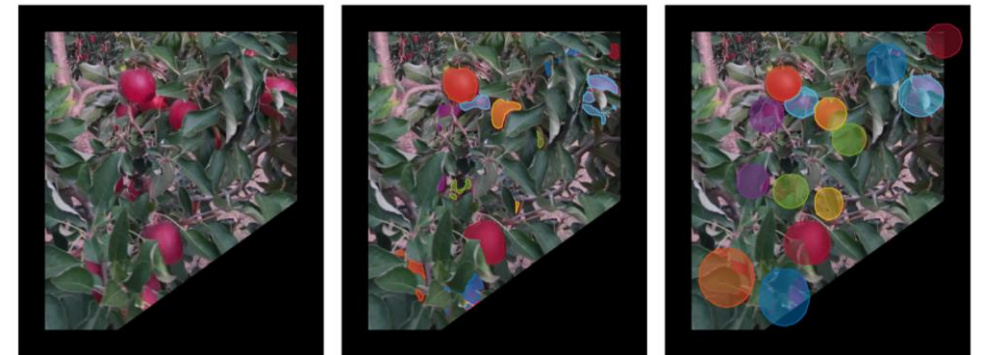
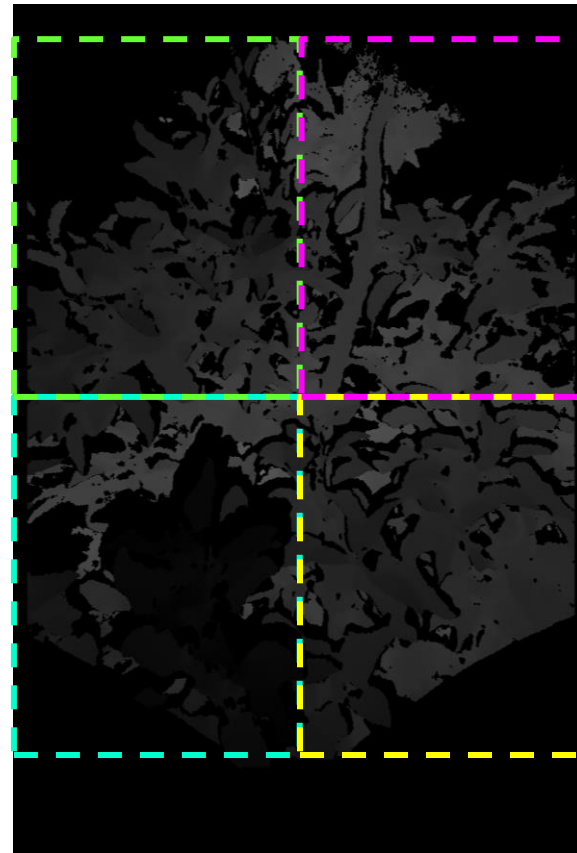
3925 RGB-D images annotated with:

- Modal segmentations masks
- Amodal segmentation masks
- Diameter ground truth

Fruit sizing

Experimental set-up

Story apple orchard data (Used for FruitsAlz evaluation)

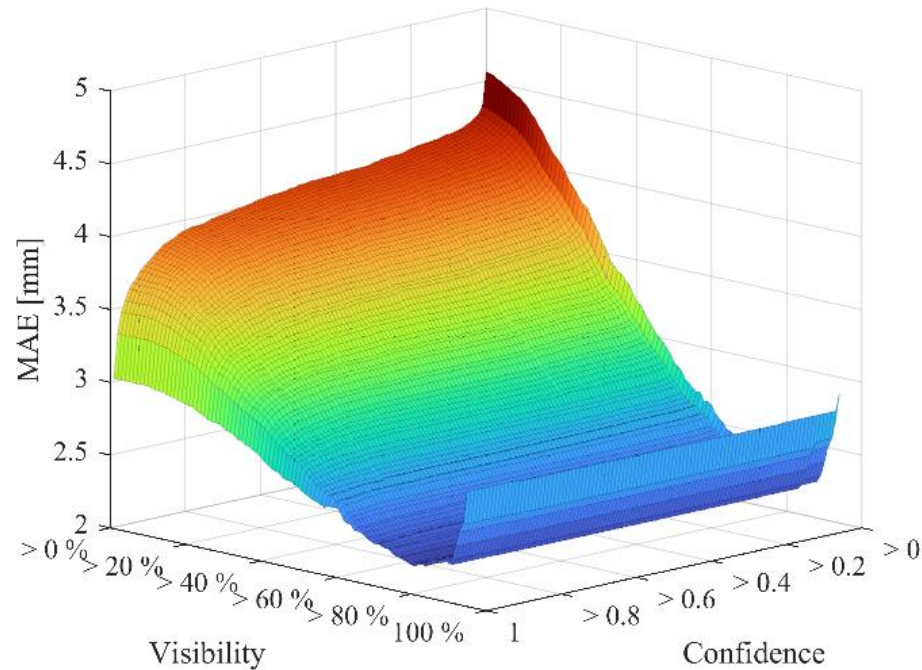


Mean size (mm)

| | | | |
|----|----|----|----|
| 73 | 72 | 73 | 73 |
| 77 | 69 | 71 | 70 |
| 77 | 72 | 73 | 70 |
| 77 | | 74 | |
| 75 | | 75 | 75 |
| 77 | | 72 | 72 |
| 72 | 69 | 73 | 73 |
| 76 | 74 | 74 | 77 |
| 72 | 76 | 72 | 75 |
| 71 | 74 | 76 | 74 |
| 79 | 75 | 74 | 78 |
| 78 | 77 | 77 | 74 |
| 75 | 77 | 81 | 78 |
| 75 | 75 | 76 | 77 |
| 76 | 72 | 73 | 75 |
| 76 | 73 | 72 | 74 |
| 77 | 76 | 73 | 74 |
| 77 | 76 | 72 | 75 |
| 75 | 75 | 73 | 76 |
| 76 | 72 | 69 | 72 |
| 75 | 69 | 74 | 75 |

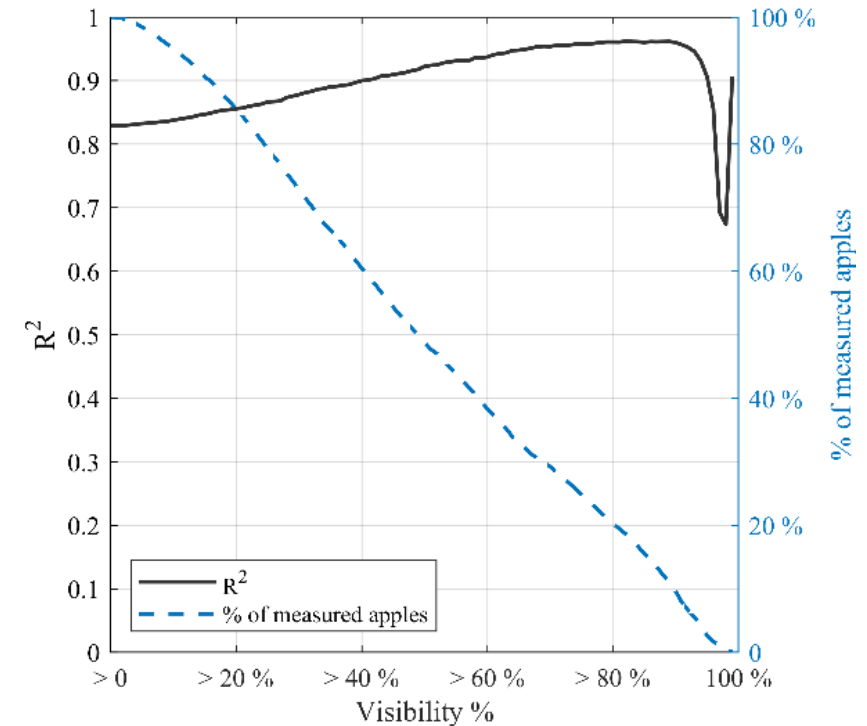
Results

Evaluation on Papple RGB-D-Size dataset

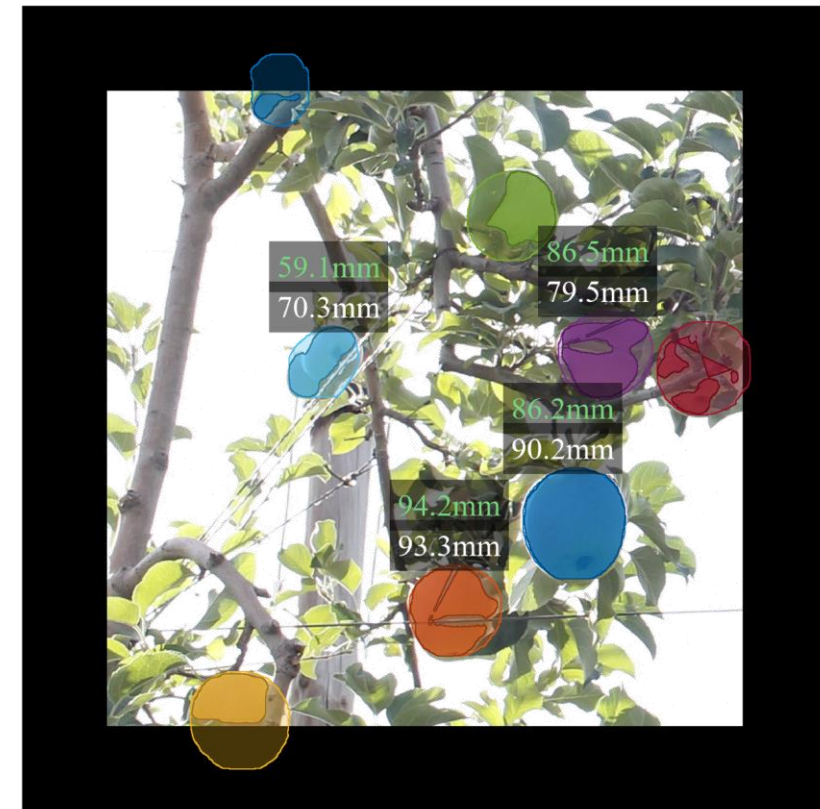
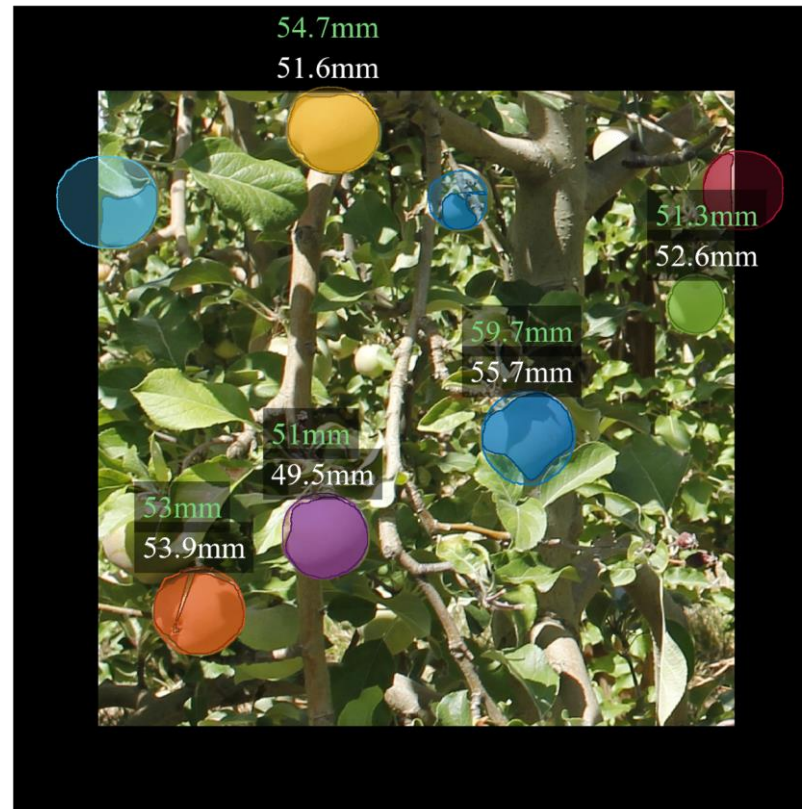
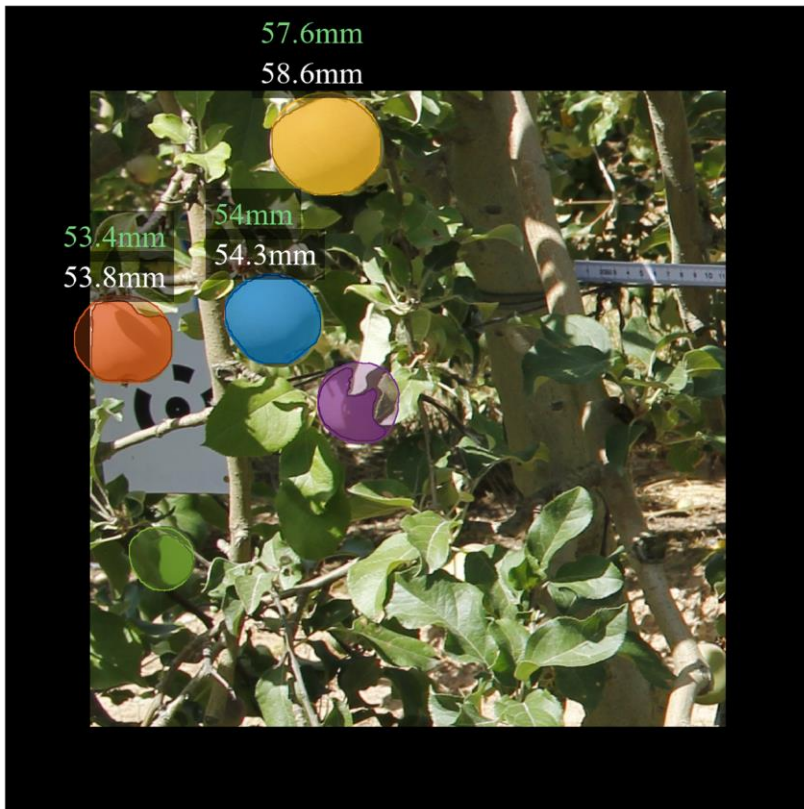


Evaluation of apples detected with a Visibility >60%

| <i>MAE</i> (mm) | <i>MBE</i> (mm) | <i>MAPE</i> (%) | <i>RMSE</i> (mm) | <i>R</i> ² |
|--------------------|--------------------|--------------------|---------------------|-----------------------|
| 2.93 | -0.19 | 4.19 % | 4.14 | 0.91 |



Results



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Fruit Counting

- ByteTrack consistently demonstrated slightly higher performance compared to SORT, while DeepSORT exhibited lower scores.
- Fruit counting task showed reduced errors and improved performance closer to the harvest date.
- Results showed accurate yield predictions with MAPE = 7.47 % and $R^2 = 0.7$.
- The qualitative evaluation showed the system's ability to identify zones of higher and lower fruit production in the orchard.

Fruit Sizing

- The CNN was able to robustly estimate the actual shape of fruits that were partially occluded.
- The simultaneous prediction of modal and amodal masks allow to estimate the percentage of occlusion of each detected apple.
- The estimated visibility allowed to discard the measurement of highly occluded apples.
- Future works will extend this evaluation to measure the size of other fruit varieties and fruitlets.

Acknowledgements

This work was partly funded by the Departament de Recerca i Universitats de la Generalitat de Catalunya (grant 2021 LLAV 00088), the Spanish Ministry of Science, Innovation and Universities (grants RTI2018-094222-B-I00 [PAgFRUIT project], PID2021-126648OB-I00 [PAgPROTECT project] and PID2020-117142GB-I00 [DeeLight project] by MCIN/AEI/10.13039/501100011033 and by “ERDF, a way of making Europe”, by the European Union).



FruitsAlz

Research Group in AgrolCT & Precision Agriculture



Universitat
de Lleida



UNIVERSITAT POLITÈCNICA
DE CATALUNYA
BARCELONATECH



Institució
CERCA
Centres de Recerca
de Catalunya

Jordi Gené Mola

Efficient Use of Water in Agriculture Program,
Institute of AgriFood, Research and Technology
(IRTA).

Departament de Ciència i Enginyeria Forestal i
Agrícola. Universitat de Lleida – Agrotecnio.

E-mail: jordi.gene@irta.cat