New Technologies for Plant Phenotyping Unidad Integrada Balcarce (INTA-UNMDP) 4 de mayo de 2016

Low cost computer vision implementations for plant phenotyping/identification problems

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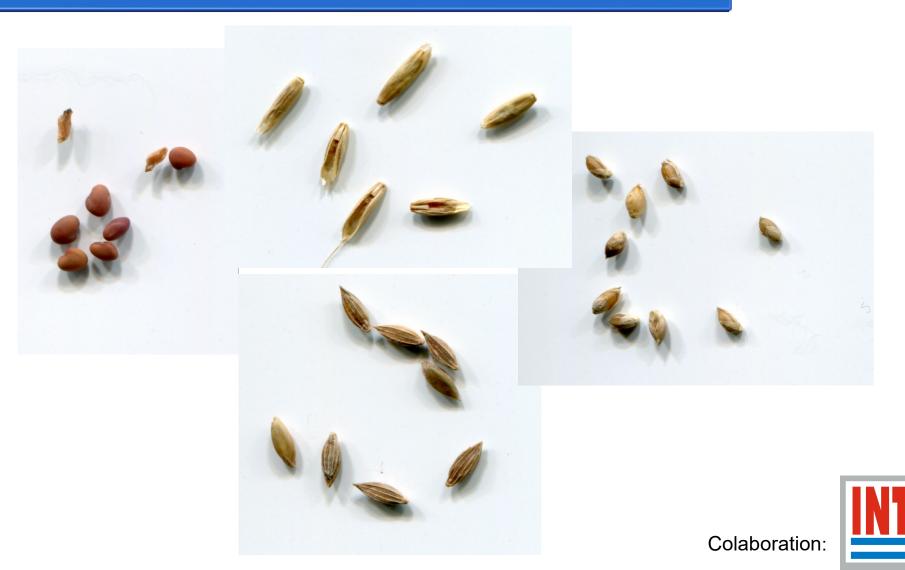
Outline



- Our path here:
 - Weed seeds
 - Green seeds
 - Plant identification using veins
 - Counting seeds in pods
 - Stripes in apples
- Conclusions
- The Future

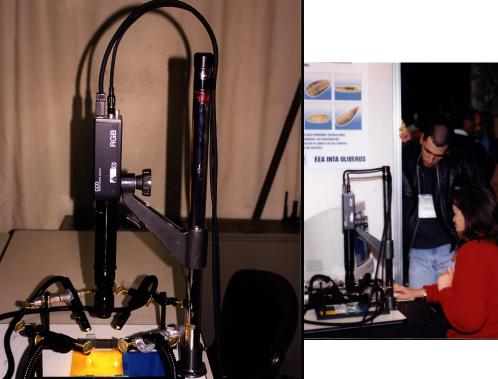
The beginning: Weed seeds identification (~2000)





Weed seeds identification: Hardware

- High-End Equipment
 - Frame grabber
 - Special camera
 - Light source
 - Etc.
- Pro: High Performance
- Con:High cost!

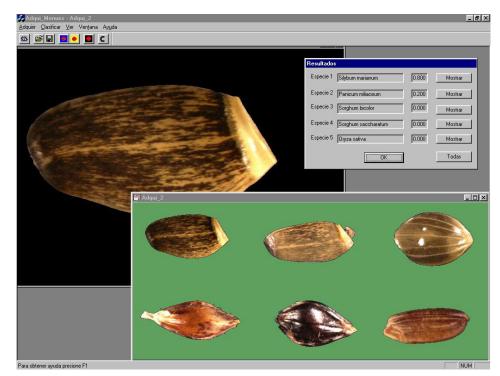




Weed seeds identification: Software



- Measurement of diverse features:
 - Morphological
 - Textural
 - Color
- Classification with Neural Networks ensembles
- Very good results:
 - +95% correct recognition on 250 species
 - +99.5 accuracy using the 5 most probable species





Weed seeds identification: The problems



- Nobody was willing to pay the cost of the equipment!
- High-End video equipment also have problems
 - Drivers
 - Replacements
 - Aging of lamps (COLOR!)

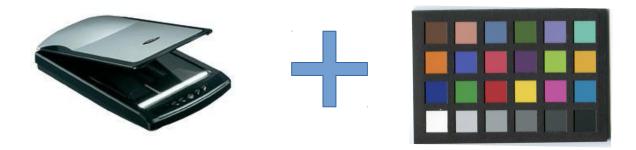
Second attemp: Green levels in soybeans (~2008)





Green levels in soybeans: How to measure color?

- We gave up on special hardware!
- Low cost solution:
 - Of-the-shelf imaging device with calibration standard
 - Software implemented as a web service





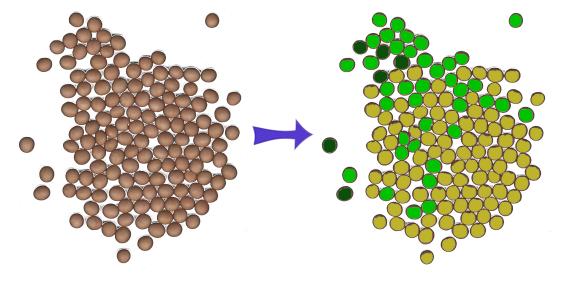


Green levels in soybeans: Software



- Calibrated Scanner + Segmentation
- Feature extraction
 - Morphological
 - Color
- Clasification with Random Forest (Ensemble of classification trees)

 All project based on Open Software (Open CV - R)



Green levels in soybeans: Problems!

- Color is really difficult!
- Even for us!
- We can control the ilumination easily with a flatbed scanner, but translating colours from diverse equipments with high accuracy is very difficult



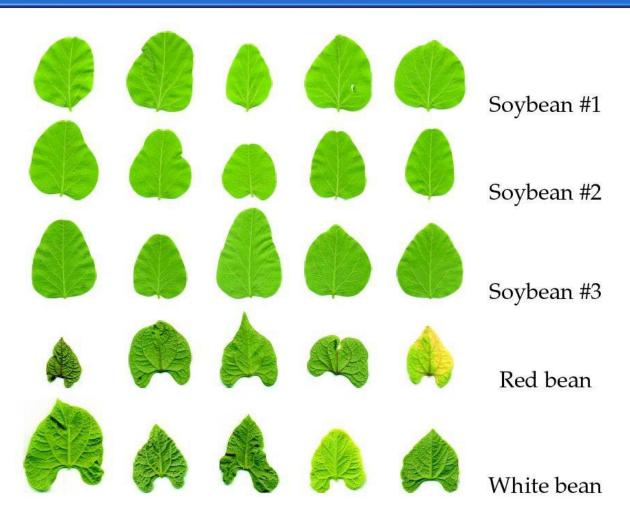


Green levels in soybeans: Results



- Average human accuracy: 65%
- Best result for automatic system: 85%
- But:
 - Using a single scanner
 - Translating from other scanners decrease accuracy to near random results

Cultivar identification using leaf veins (2012)





Colaboration:

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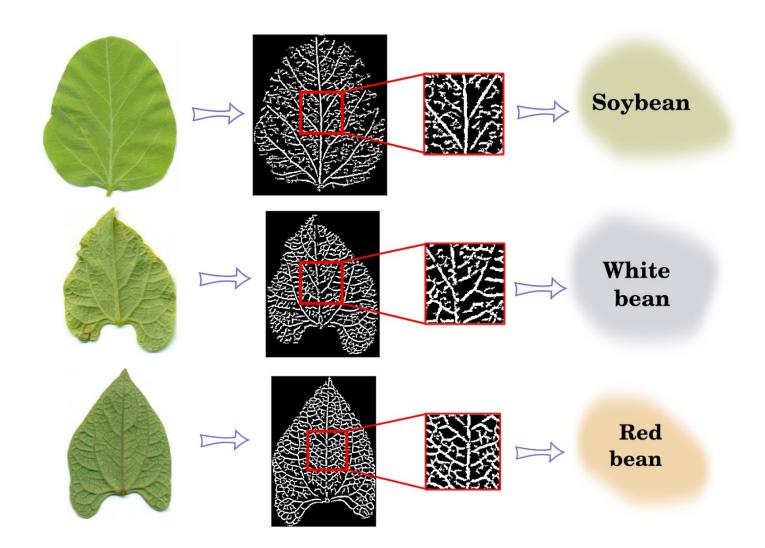
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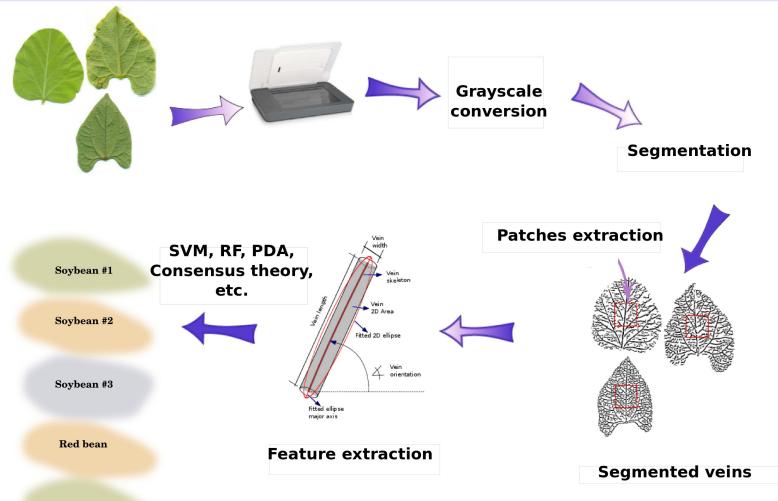
Cultivar identification using leaf veins





Cultivar identification using leaf veins: pipeline





White bean

Cultivar identification using leaf veins: results

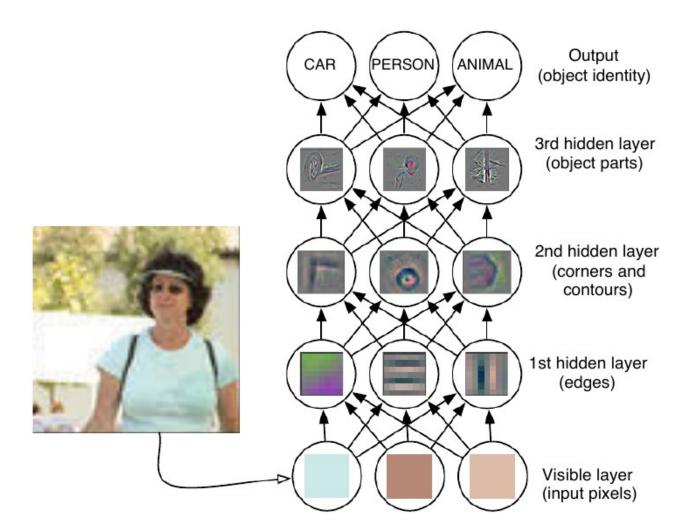


- Average human accuracy: 45%
- Best result for automatic system: 60%

- Automatic methods outperforms humans (on cultivars and species)
- But results are not good enought as to develop a portable device

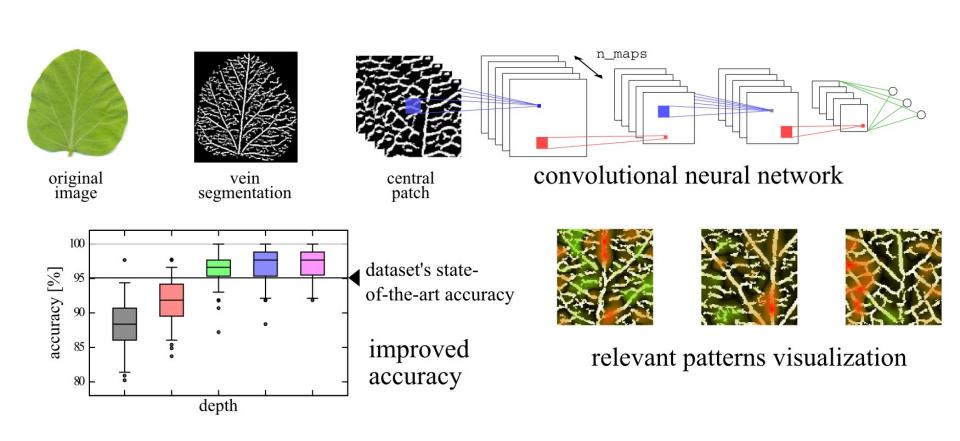
Cultivar identification: can we improve? Deep learning





Cultivar identification with Deep learning





Phenotyping: counting seeds in pods (2015)

- Semi-automatic procedure: pods are colected from the plant by hand and counted automaticaly with a vision system
- Regular camera, cheap ilumination device and a computer
- Segmentation + feature extraction
- Classification with SVM

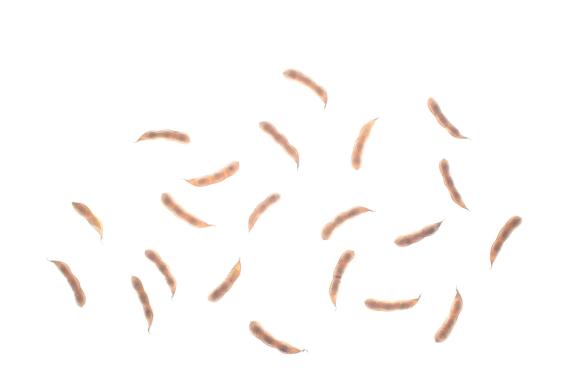






Phenotyping: counting seeds in pods

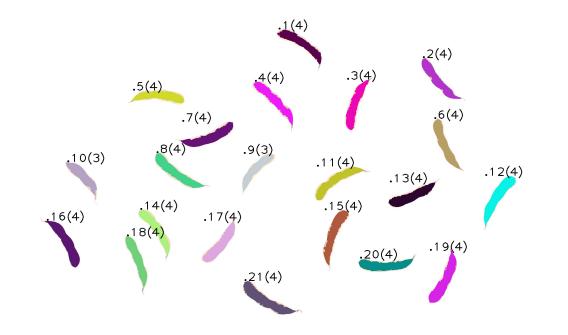




Phenotyping: counting seeds in pods

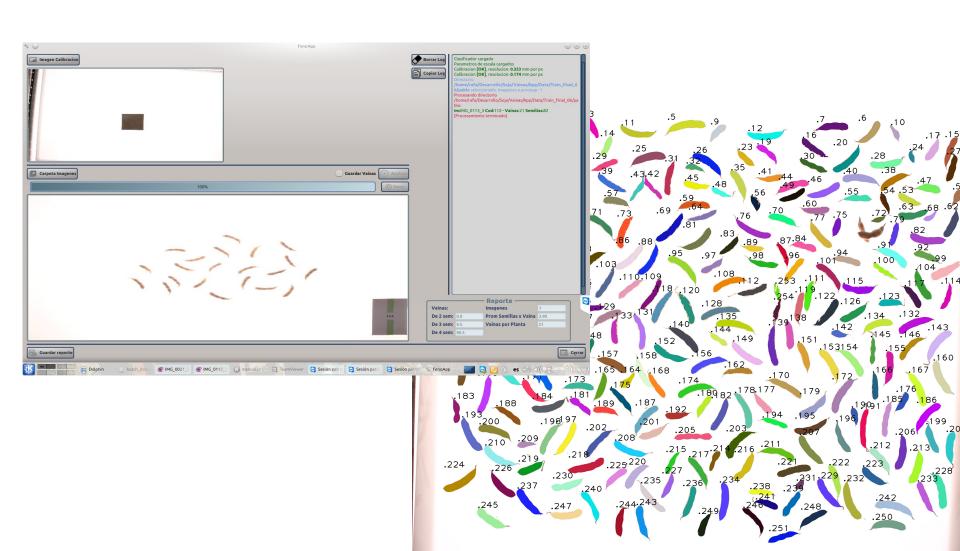


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Phenotyping: counting seeds in pods

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Phenotyping: Results



- Accuracy +90%
- Limits: pods with "new" shapes and size lead to errors
- Proposed solution: using deep learning (working now...)

Phenotyping: stripes on apples (2015)









- Work in progress with FEM (Trento, Italy)
- Goal: develop a low cost device to grade apples according to stripes quality

Conclusions



- Machine vision systems based on low cost hardware are useful and easy to develop
- Many agricultural applications known
- Measuring color in practice is difficult
 - But you hardly need color in phenotyping
- Lots of potential phenotyping applications

The (near) future



- Phenotyping
 - Counting seeds (pods) in live plant
- Identification
 - Identifying weeds in real time video
 - Collaboration in the development of a weed control autonomous robot

The team



- Dra. Mónica Larese
- Dr. Rafael Namías
- Dr. Pablo Verdes
- Dr. Guillermo Grinblat
- Dr. Lucas Uzal
- Dr. Ariel Baya
- Dra. Belén Bernini (former)
- Dr. Alejandro Ceccatto (former)
- Dr. Hugo Navone (former)
- Dr. Roque Craviotto and gruop (INTA OLIVEROS)
- Dr. Eligio Morandi and group (UNR Zaballa)
- Dr. Eugenio Aprea (FEM Trento Italy)