

# Plant architecture phenotyping



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CIRAD



Partially inspired from presentations of M. Saudreau, et al.

# Introduction

- Plant functioning highly depends on the spatial distribution of its organs
- As such, « 3D Plant Model » are interesting support to study the different processes ...

## ➤ **Functional-Structural Plant Model (FSPM)**

- Access to parameters/information are usually difficult to obtain experimentally (root systems, 1 tree ~ 100 leaves; 1 field ~ 100 plants)
  - Need to adapt acquisition protocol

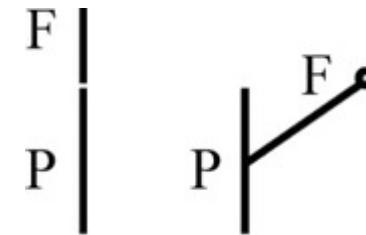
# Introduction

- Different **variables/architectural traits** with various interest ...

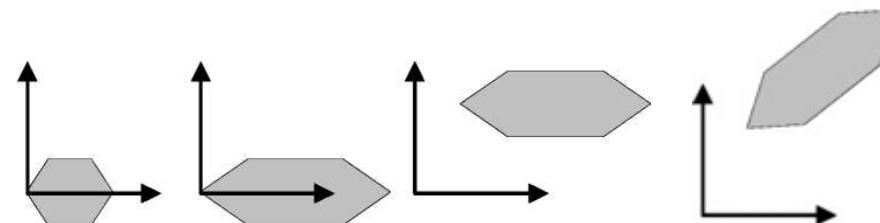
Typology - Entities



Topology - Connections



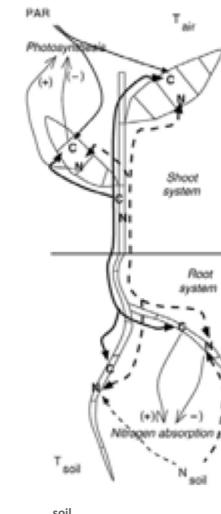
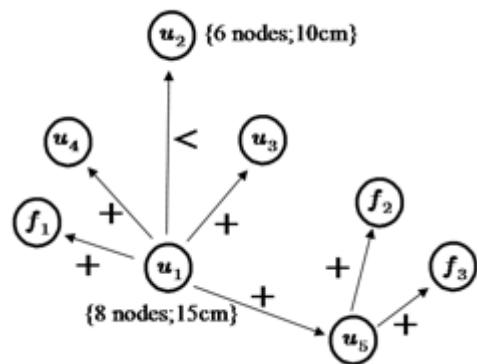
Geometry – Spatial distribution



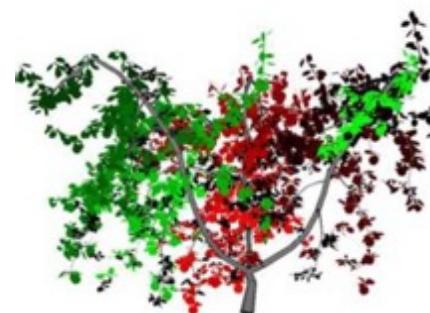
# Introduction

- Different variables/architectural traits with various interest ...

Topology

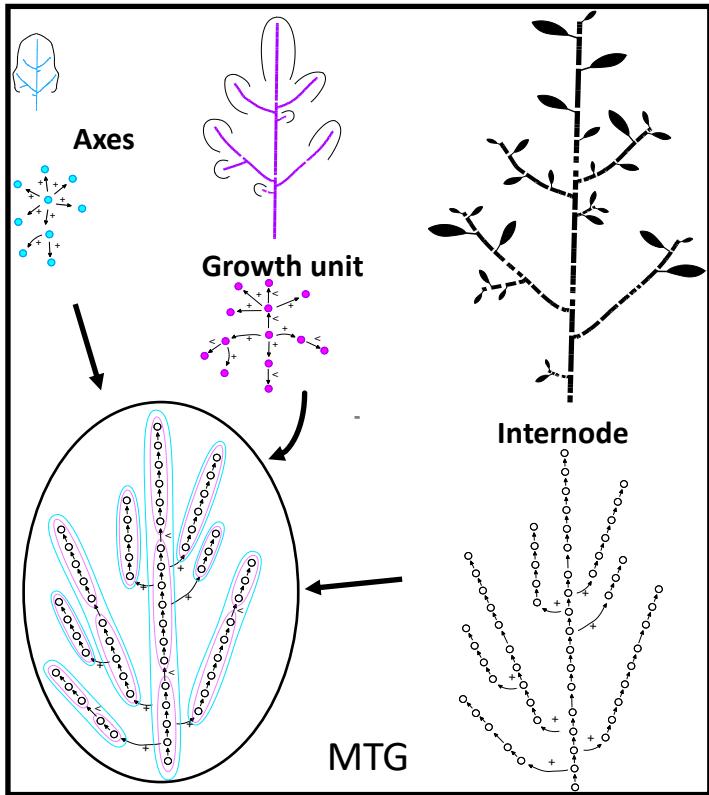


Geometry

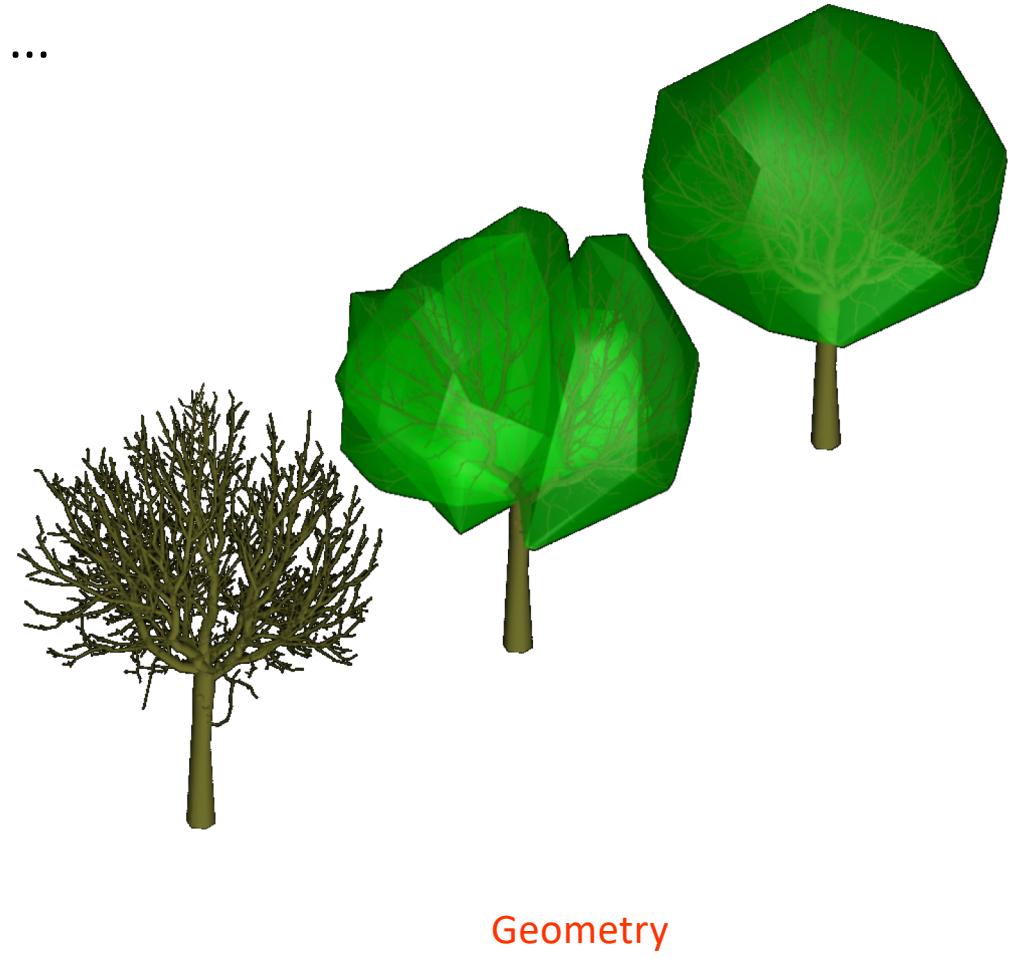


# Introduction

- Different variables/architectural traits with various interest ...



Topology

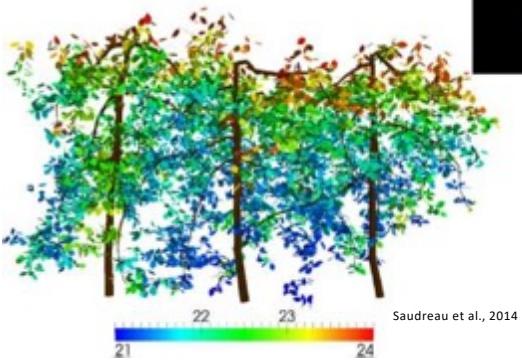


Geometry

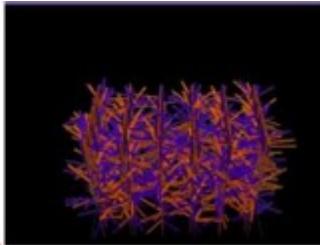
# Introduction

➤ Multiples processes/fonctions ...

3D Canopy Temperature

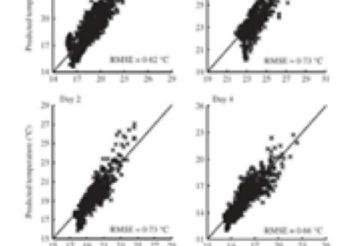
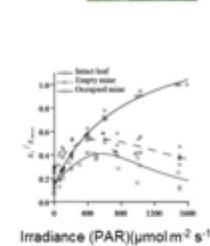
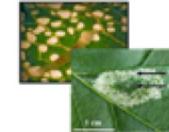


Vidal et al., 2015



Saudreau et al., 2014

Pathogen



RMSE = 0.62 °C

RMSE = 0.73 °C

Measured temperature (°C)

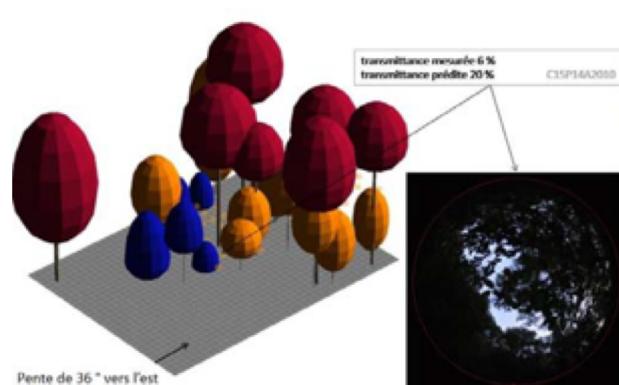
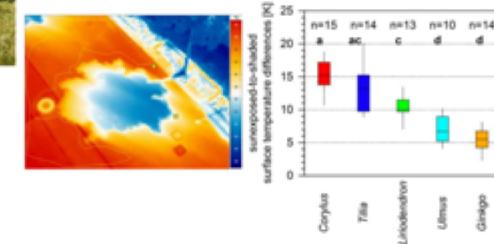
Measured temperature (°C)

Shadowing

Agroforesterie

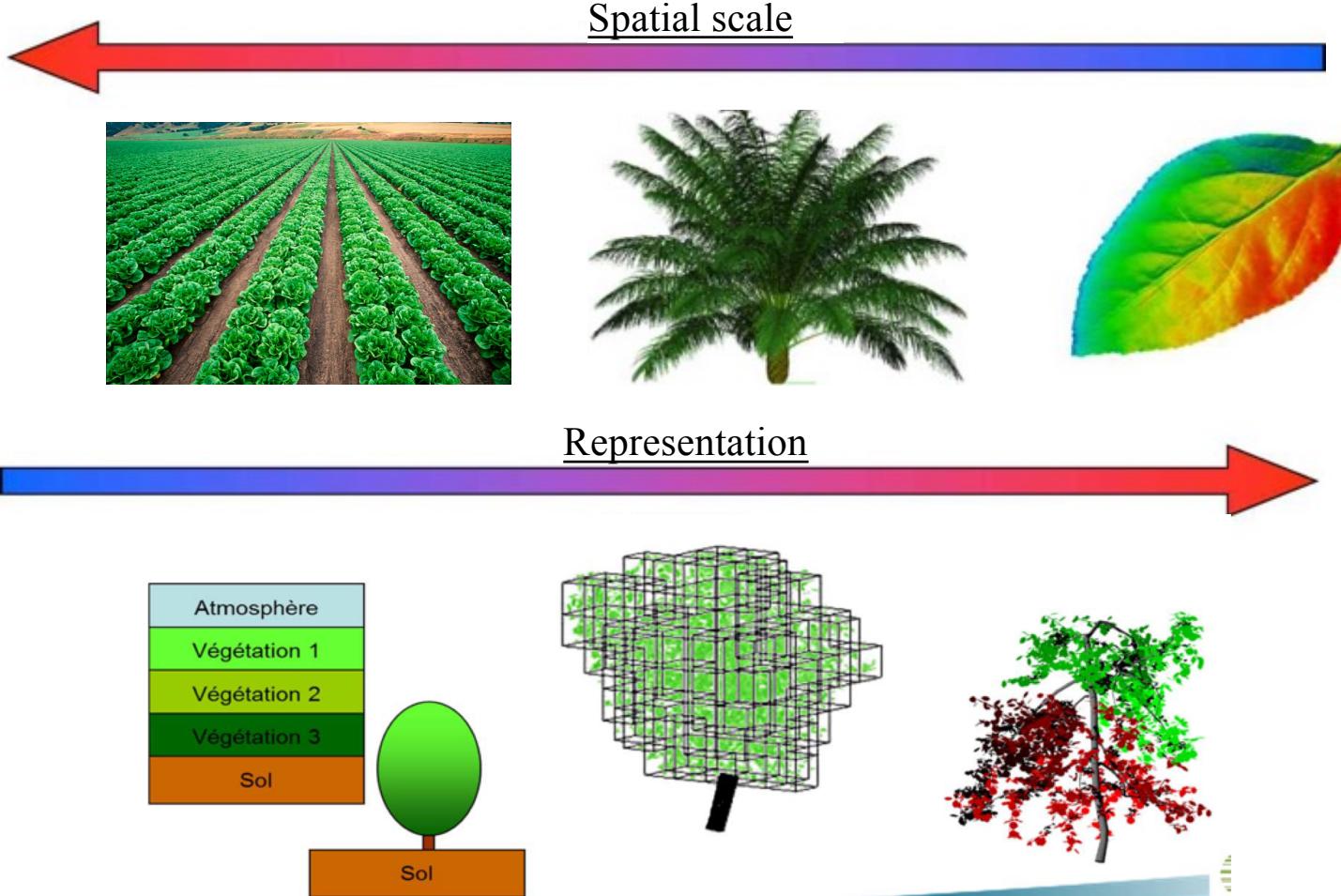


MicroClimat Urbain



# Introduction

- Variable degree of precision/details ...



- Finding the right balance between rapidity and level of precision

# Introduction

- Two approaches of methodology of acquisition coexist:

- With contacts
- Contact free



© iStockPhoto.com - esp44041882



## Contact methods



- Several «contact» methods have been developed ...



... But one was a reference for a long time



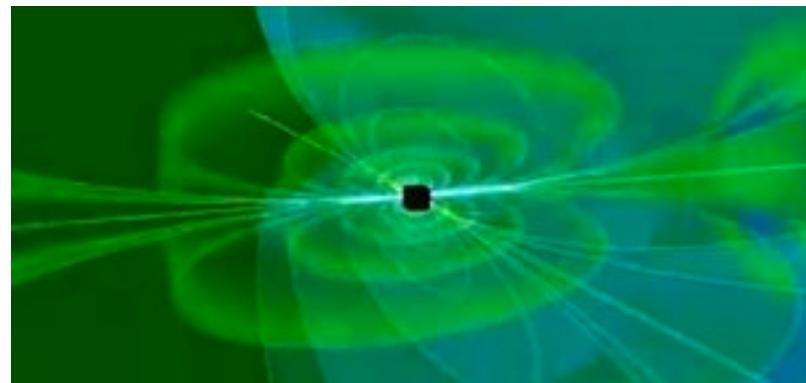
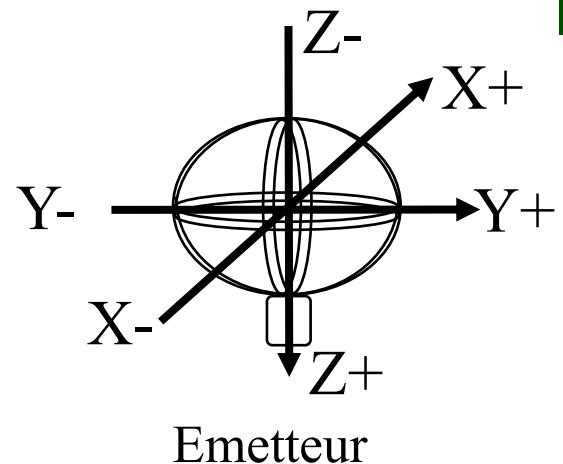
## Contact methods



- Electro magnetic fields



polhemus fastrak



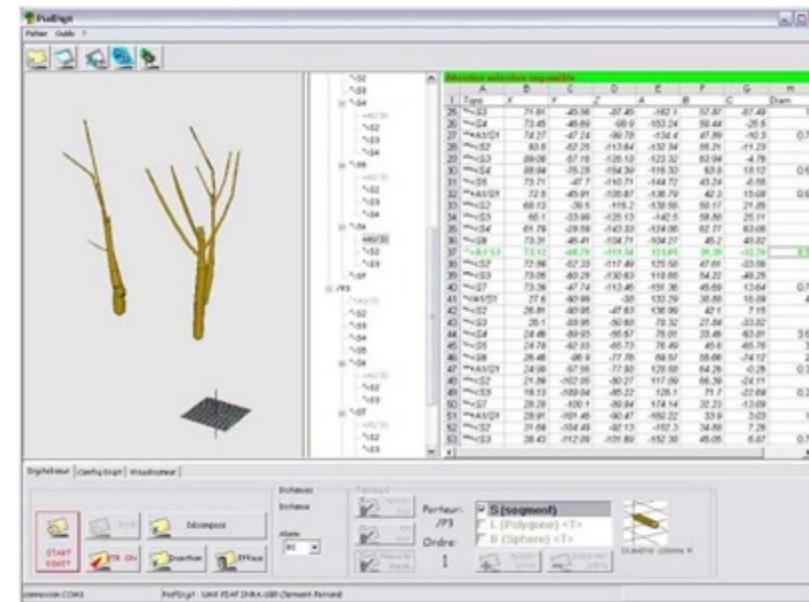


# Contact methods



A flexible protocol of acquisition (Pol95, 3A, PiafDigit);

- Simultaneous acquisition:
  - Simplified topology
  - 3D Geometry
  - Additional information
- Configurable:
  - Entities and their attributes
  - Multi-scale topological relationships
- Ergonomics:
  - Topology verified in real time
  - Real-time 3D visualization



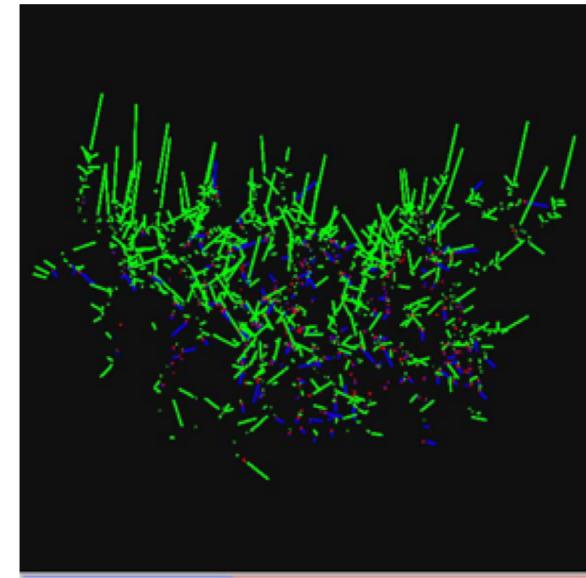
In contact with the plant, it is therefore possible to acquire additional information (type, colour, disease/insect, fruit, etc ...).



## Contact methods

- Magnetic approach:

- Semi-automatic : -
- Comprehensive list of constituents : +
- Nature of constituents : +
- Encapsulate other parameters : +



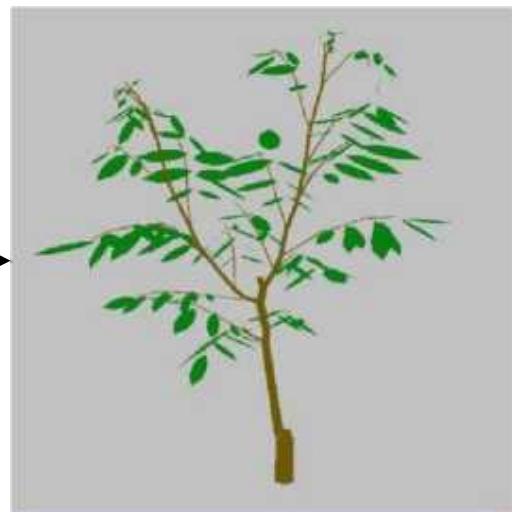
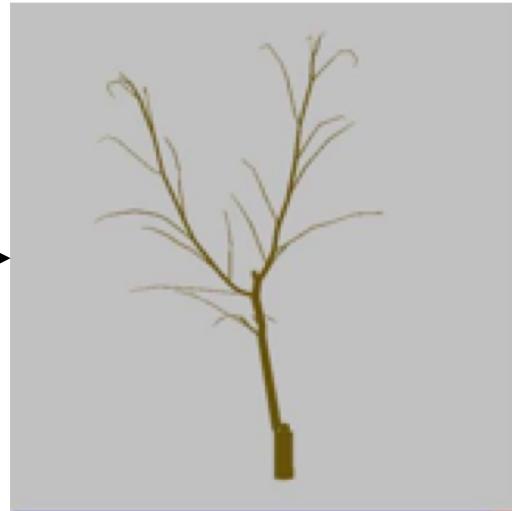


## Contact methods

Magnetic digitizing (Pol95, 3A, PiafDigit)



Reconstruction  
of young walnut



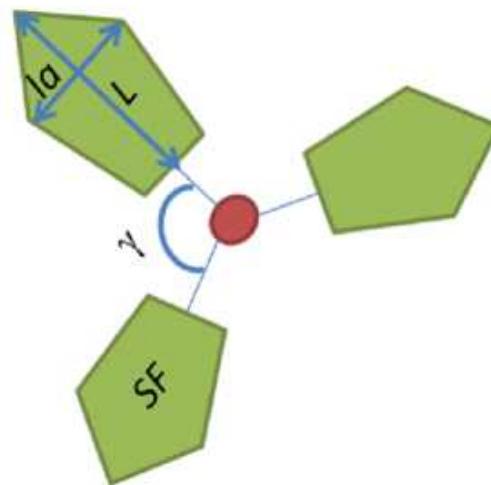
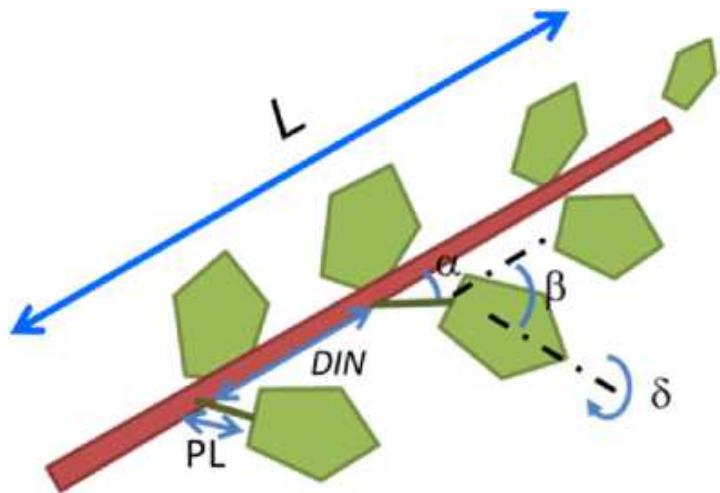
- Comprehensive but time-consuming



## Contact methods



Lighten digitizing to estimate leaf area by using **allometric relations** at shoot scale

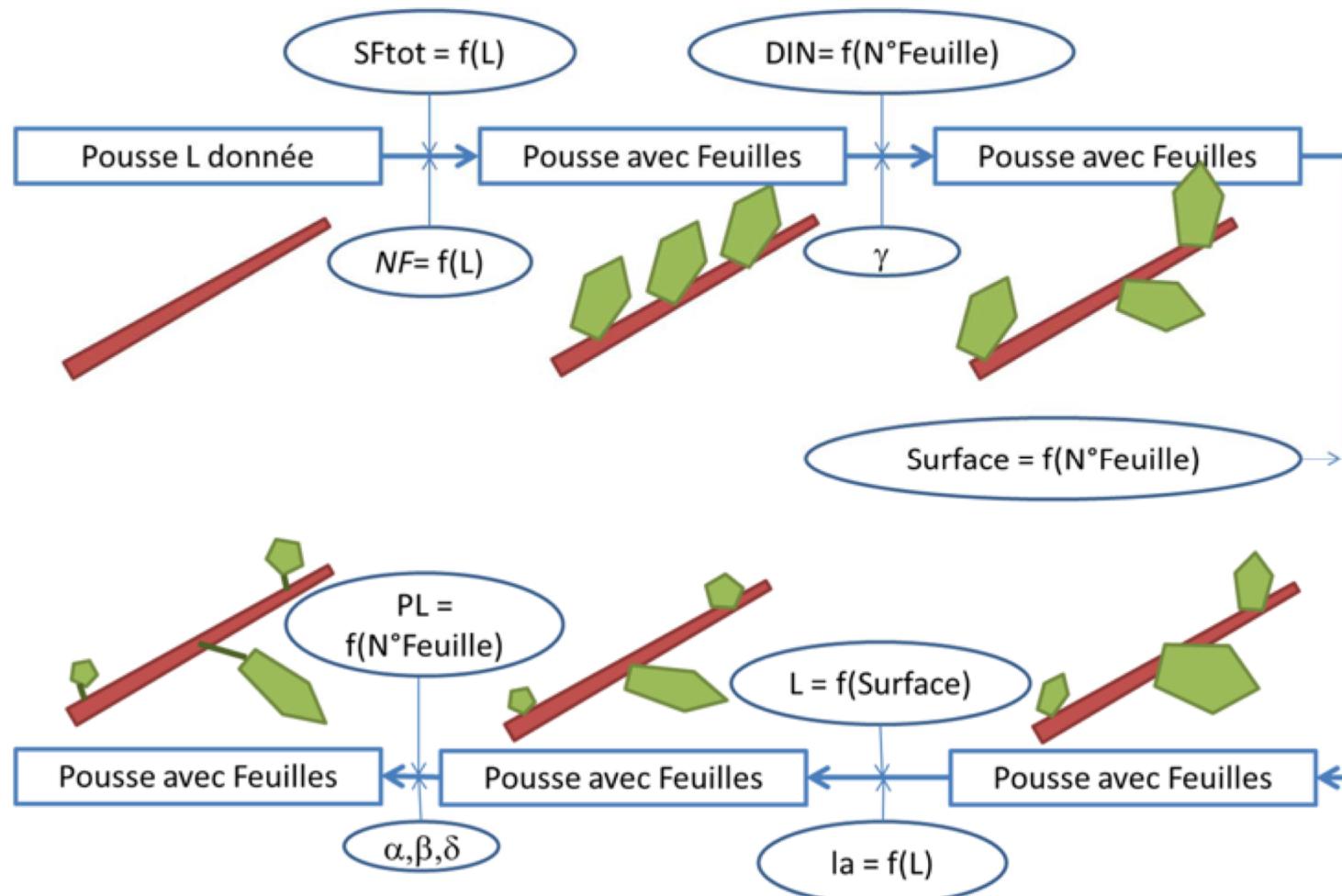




# Contact methods



Lighten digitizing to estimate leaf area by using **allometric relations** at shoot scale

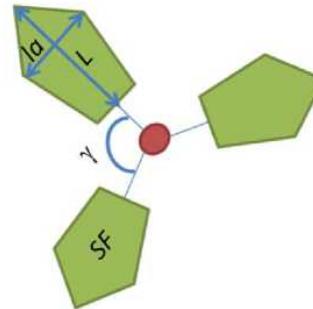
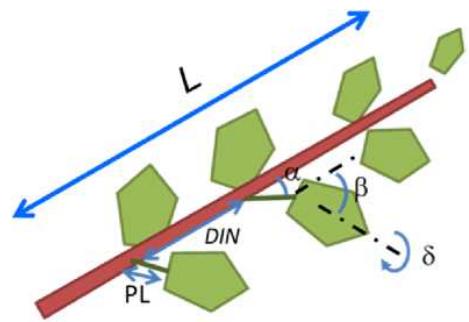




# Contact methods

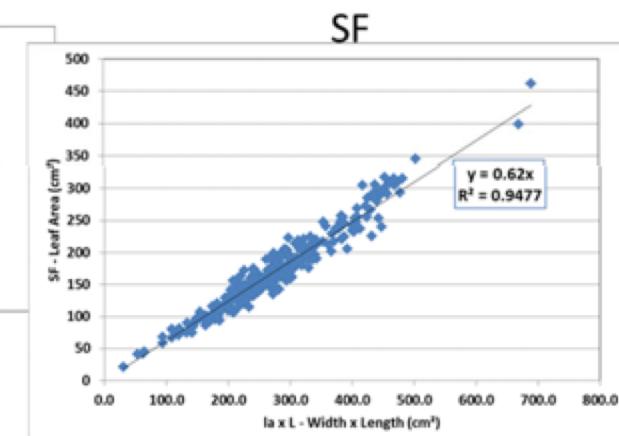
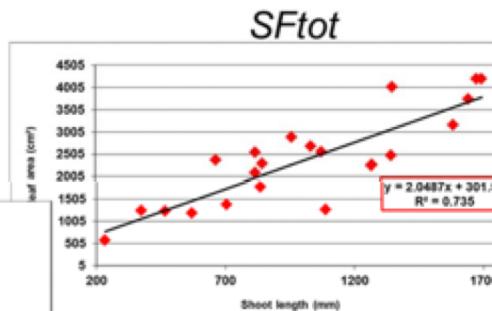
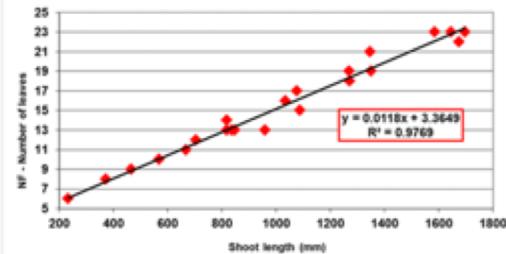


Lighten digitizing to estimate leaf area by using **allometric relations** at shoot scale



Lime tree

NF

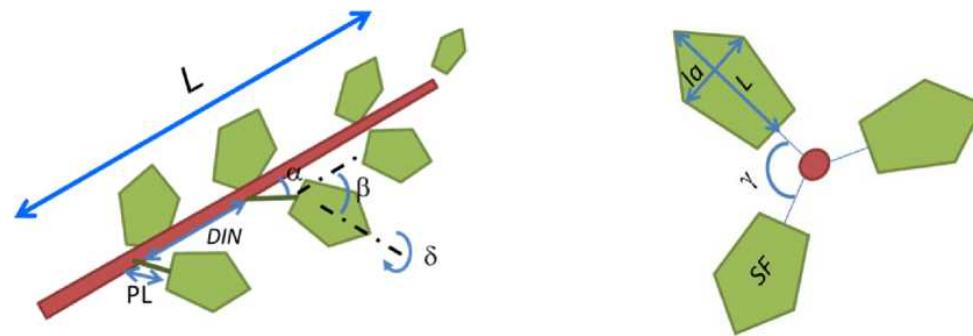




# Contact methods

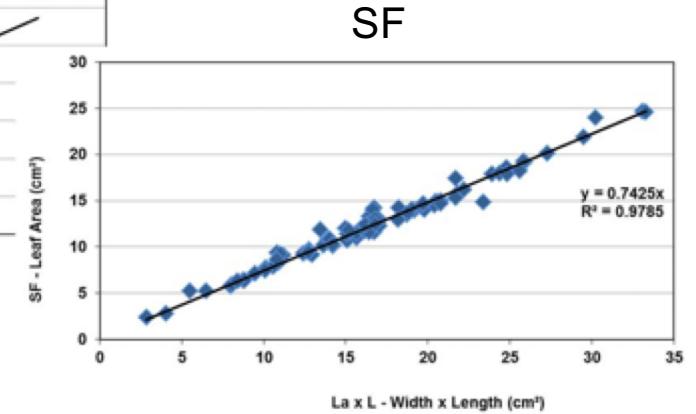
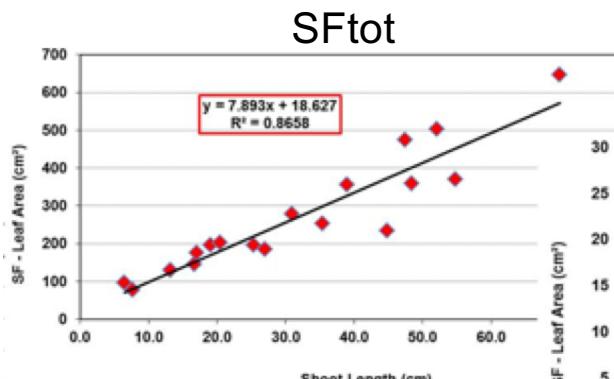
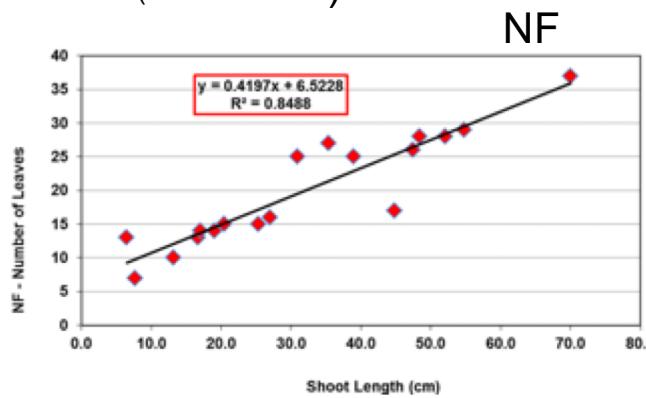


Lighten digitizing to estimate leaf area by using **allometric relations** at shoot scale



Apple tree

(cv. Pichounette)



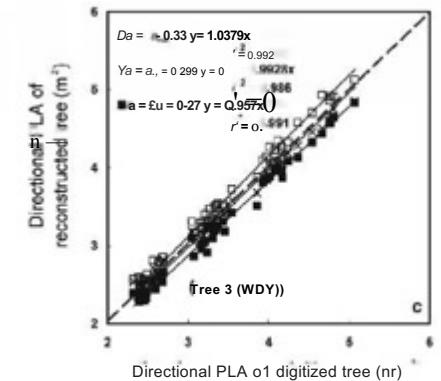
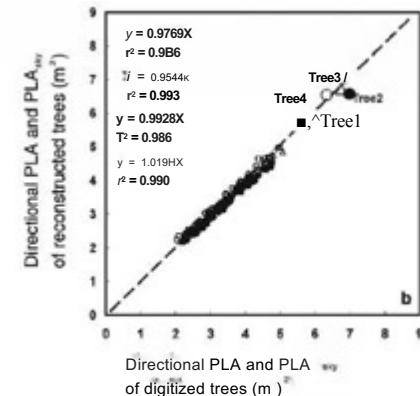
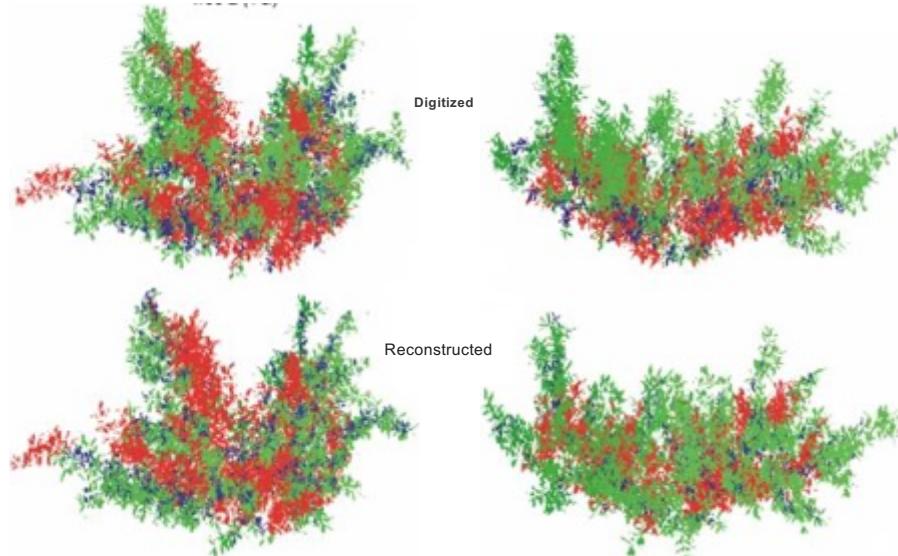


## Contact methods



Lighten digitizing to estimate leaf area by using **allometric relations** at shoot scale

Pêchers (sonohat *et al.*, 2006)





Small walnut



## Contact method

Real plants vs. Digitized plants



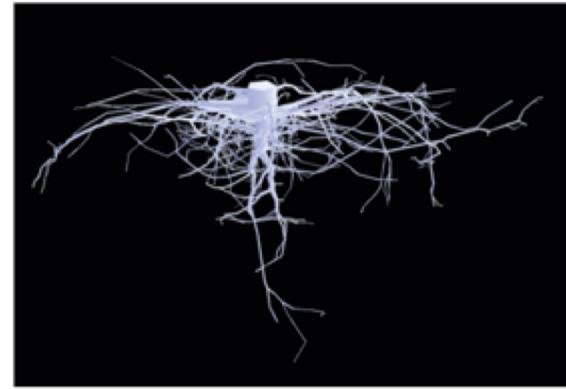
Small mango tree



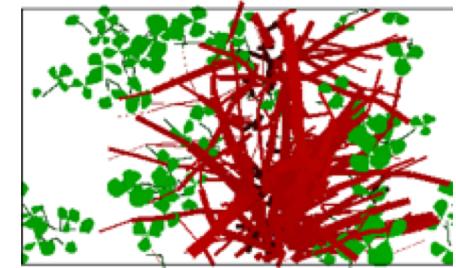
Mature walnut (8 m)



Walnut



Root system



Clover / Fetueque Mixture

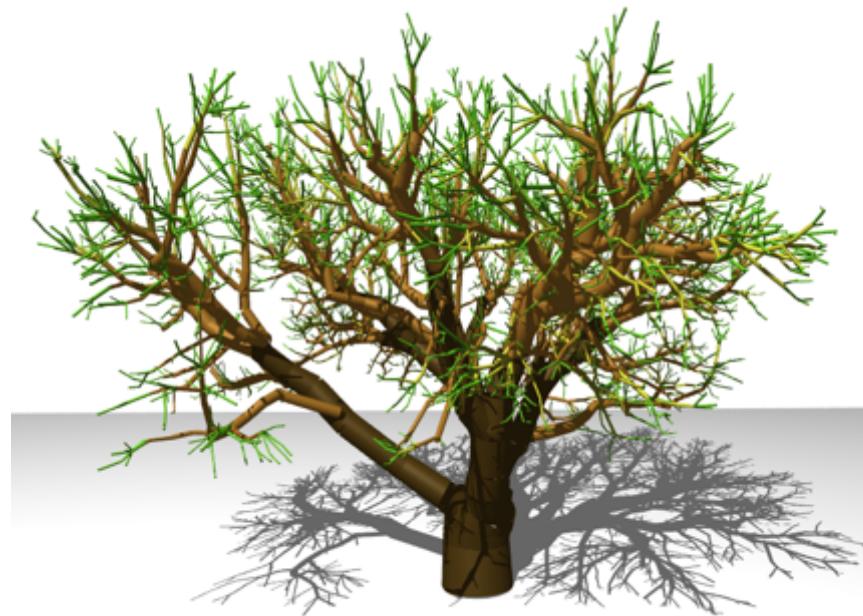


# Measuring Architecture

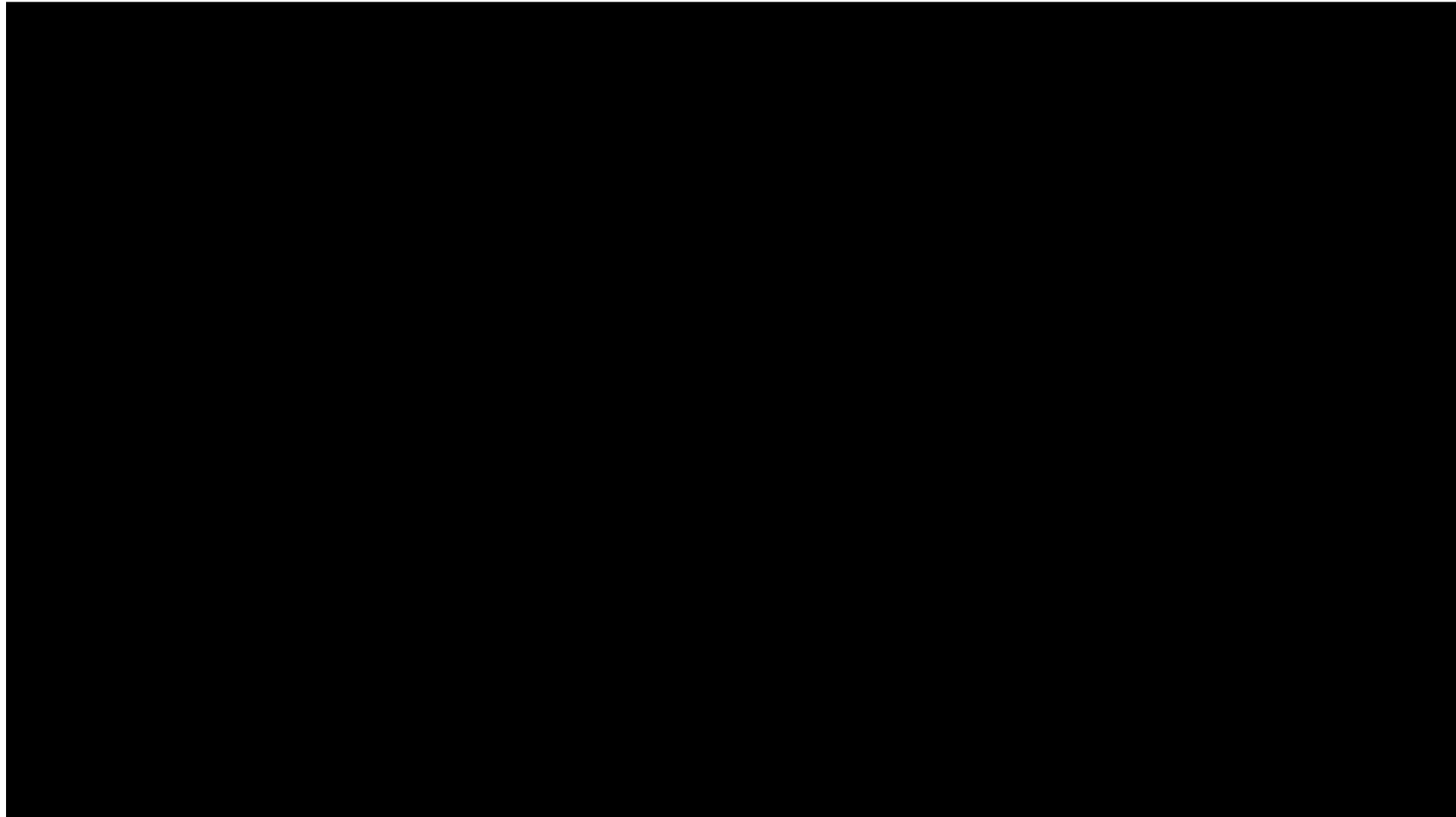
- One 13-years-old mango tree
    - For every growth unit:
      - Its top coordinates,
      - Its diameter,
      - Its type (Old, Recent),
      - The number of leaves,
      - ...
    - Around 15000 clicks for approx. 8000 growth units
- Measured in 1 month.



# Results



# Results



Participation of A. Marquier, I. Grechi, S. Persello, C. Soria and F. Normand



## Contact methods

Some applications

Comparison architecture / shoots

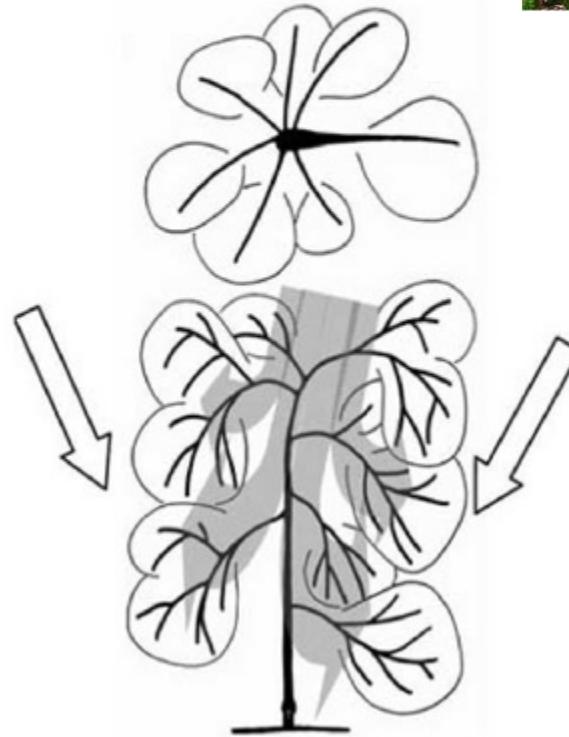


Witness Yield



Centrifugal Yield

Centrifugal Yield



Architecture more « efficient » in terms of intercepted light ?

*Willaume et al., 2004, Trees*



## Contact methods

Some applications

Comparison architecture / shoots



Witness Yield



Centrifugal Yield

Measurement of allometric relationships

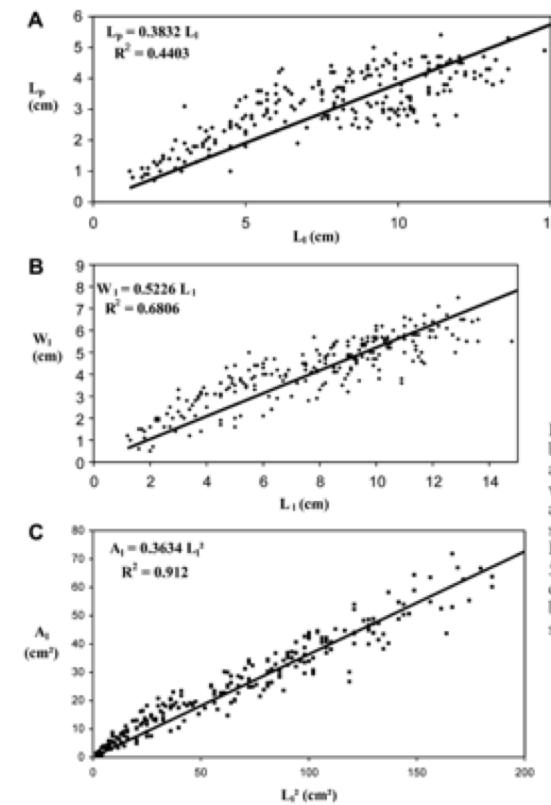


Fig. 2 Allometric relationships between a petiole length,  $L_p$ , and lamina length,  $L_l$ , **b** lamina width,  $W_l$ , and lamina length,  $L_l$ , and **c** leaf area,  $A_l$ , and the square of lamina length,  $L_l^2$ . Data collected on 230 leaves of 5-year-old "Galaxy" trees. All correlations were best adjusted by linear regressions and were significant at  $P<0.05$

Willaume et al., 2004, Trees



## Contact methods

Some applications ...

Comparison architecture / shoots

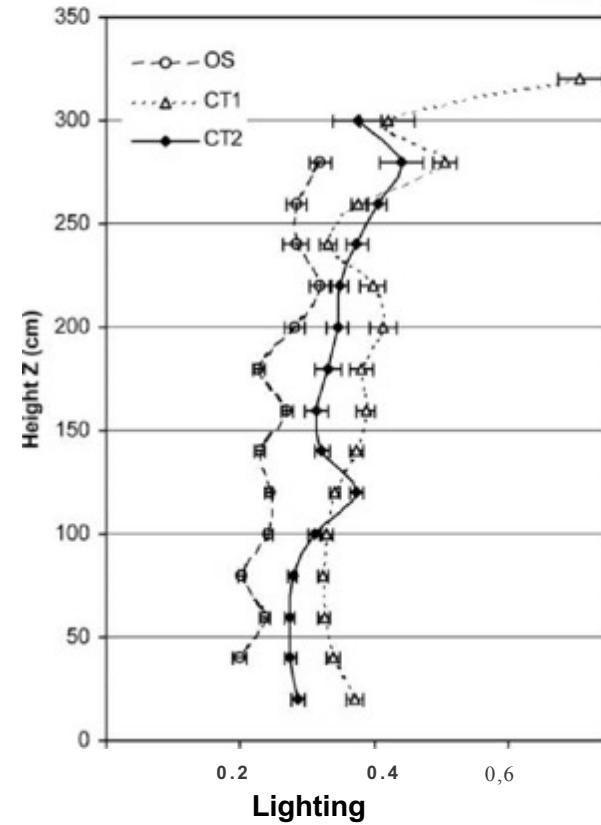


Witness Yield



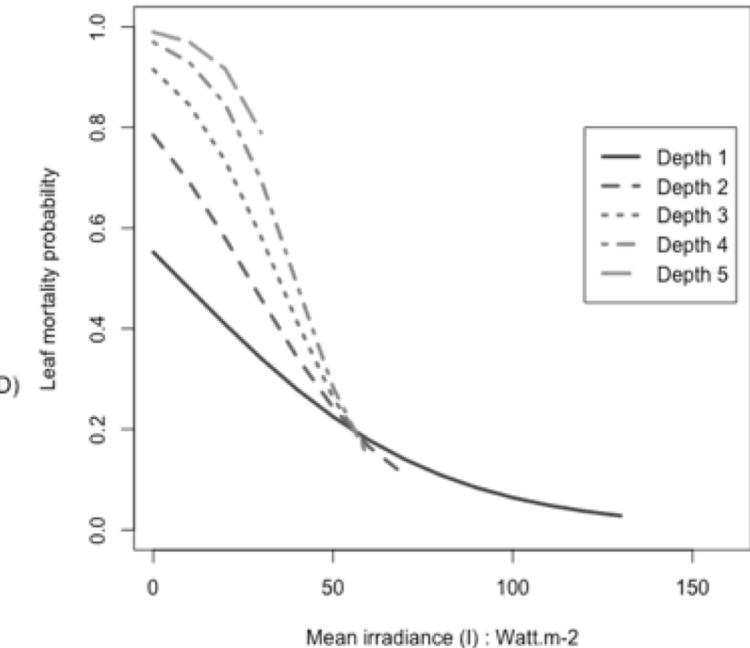
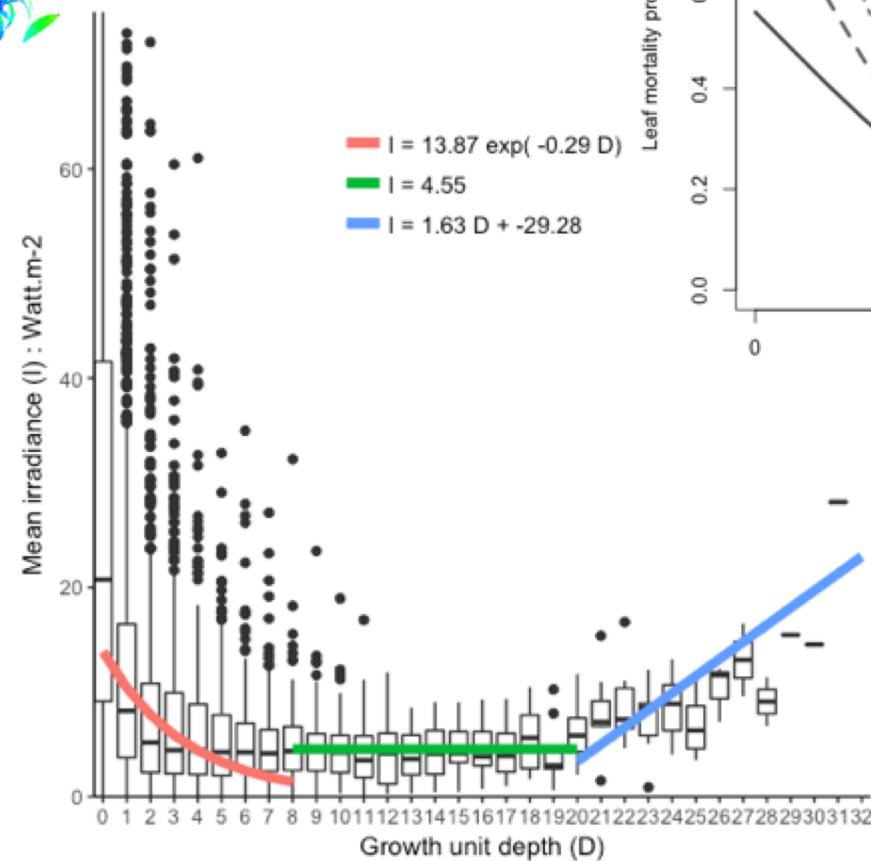
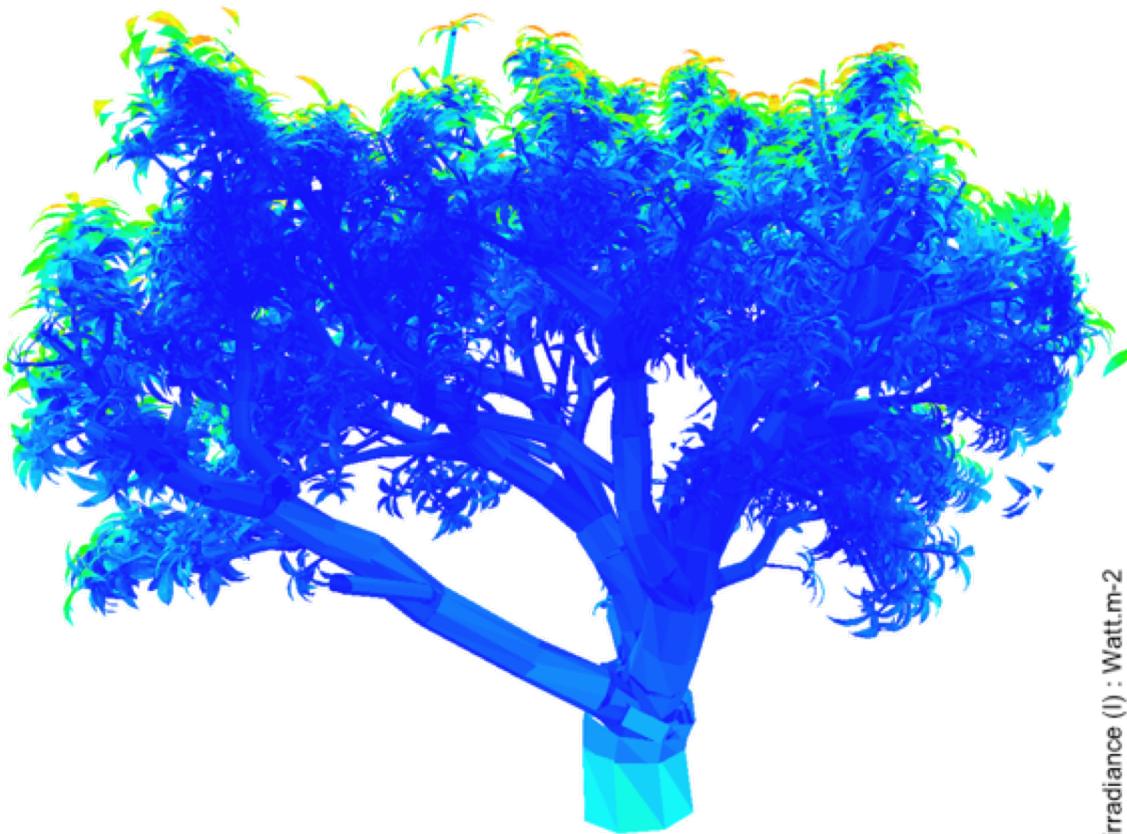
Centrifugal Yield

Comparison of intercepted light at different layers



Willaume et al., 2004, Trees

# Light interception and leaf mortality on mango trees



(Boudon et al., 2018)

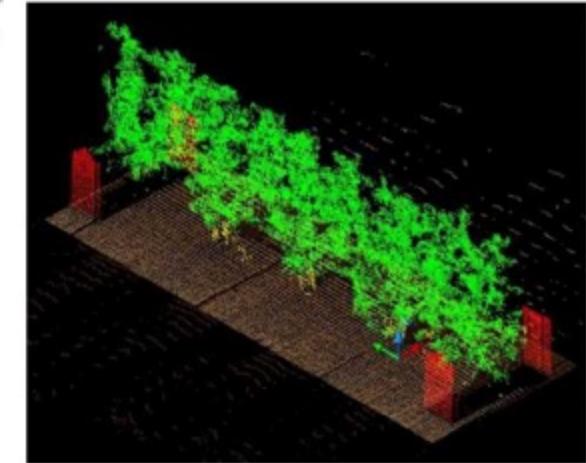
## Contactless method

Various device types for range imaging

- Registered photographs
  - Visual Hull
- Structured lighting
  - Kinect
- Photogrammetry
- Laser Scanner (or LED camera)
  - Travel time based
  - Phase shift (Interferometry) based
- Etc..



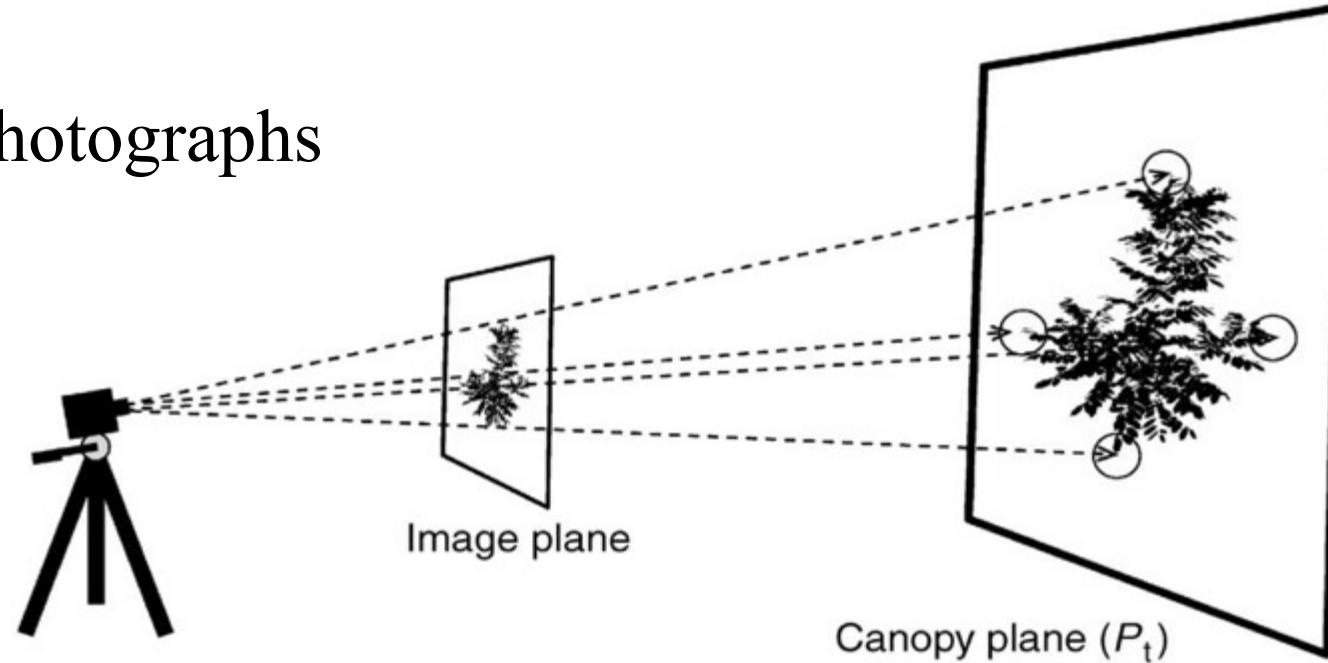
© CanStockPhoto.com - csp44041882



## Contactless method



- Registered photographs

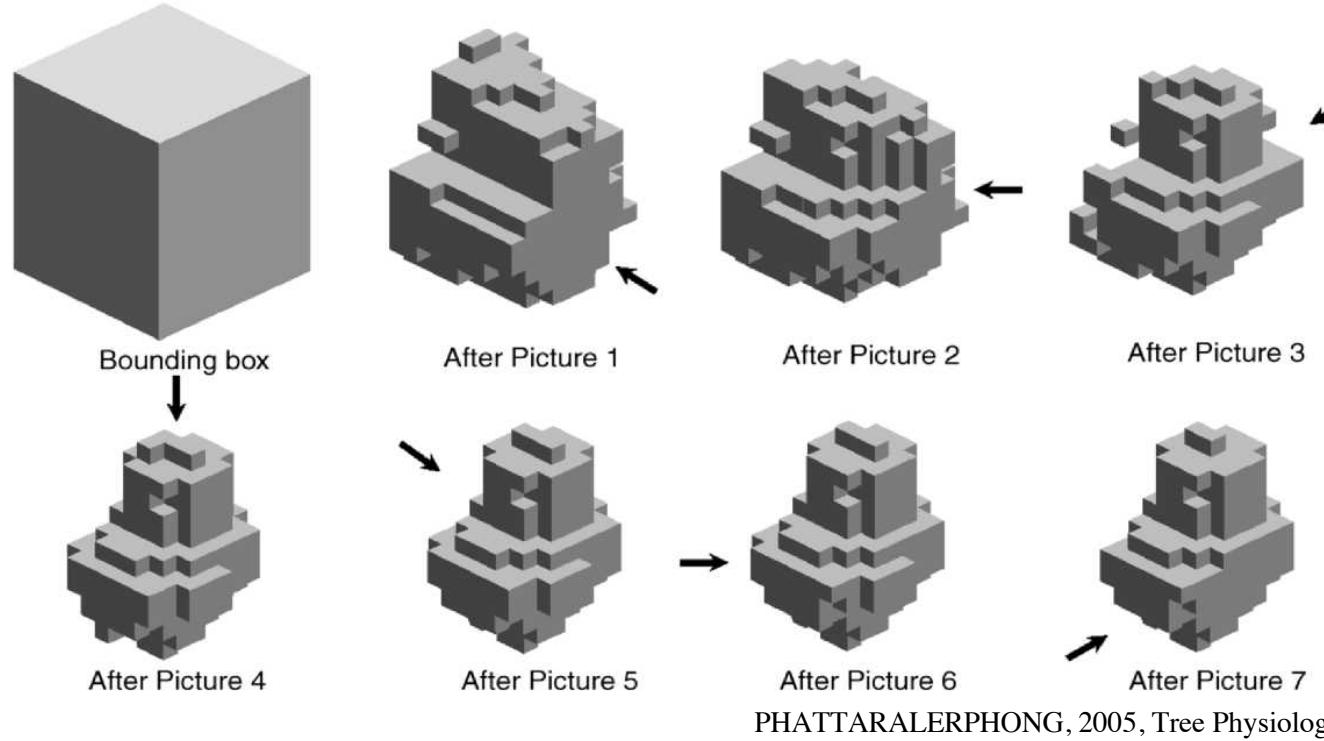


PHATTARALERPHONG, 2005, Tree Physiology

## Contactless method



- Registered photographs : space carving algorithm

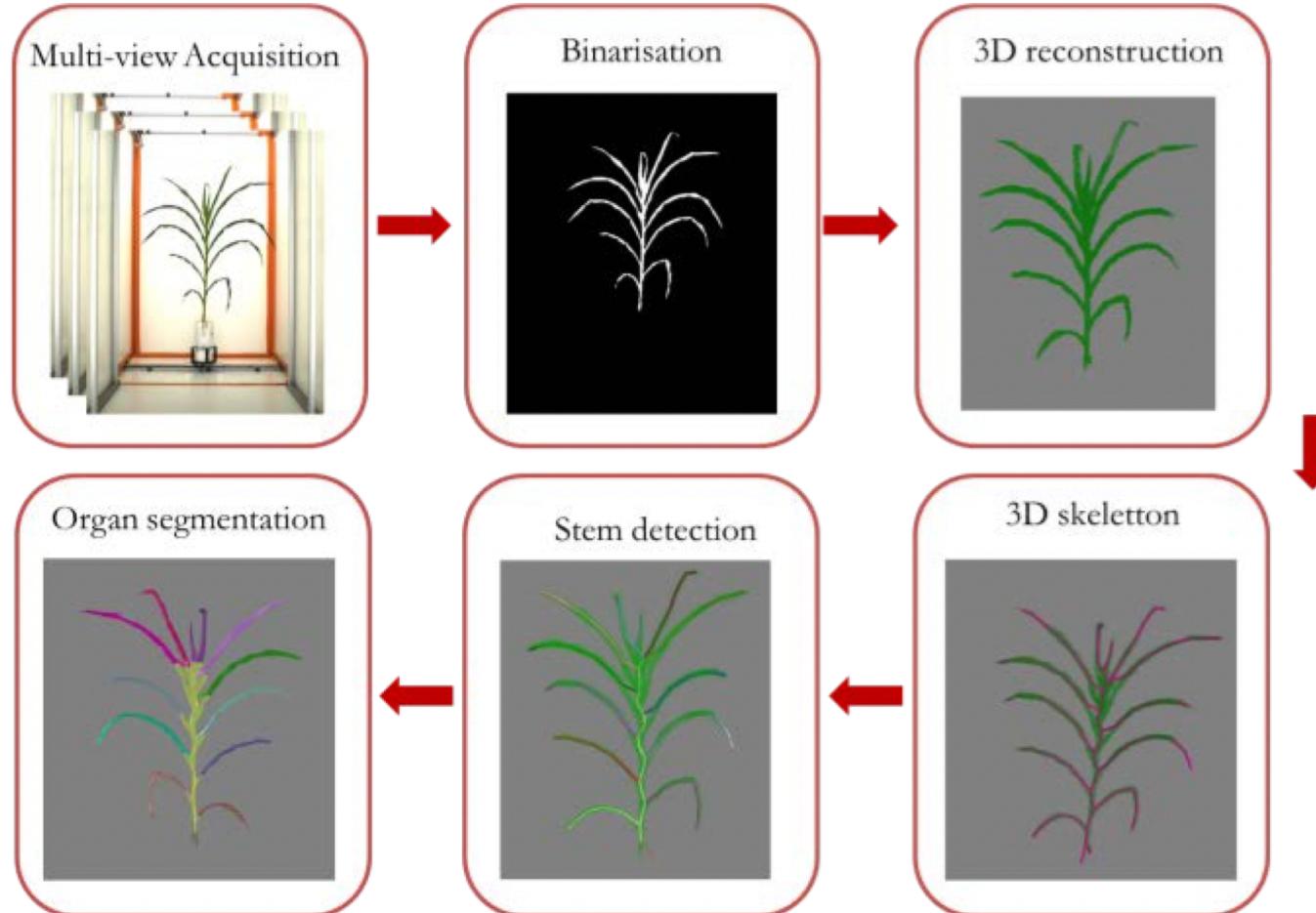


It allows to compute a volume.

## Contactless method



- Registered photographs : automated pipeline in controlled systems



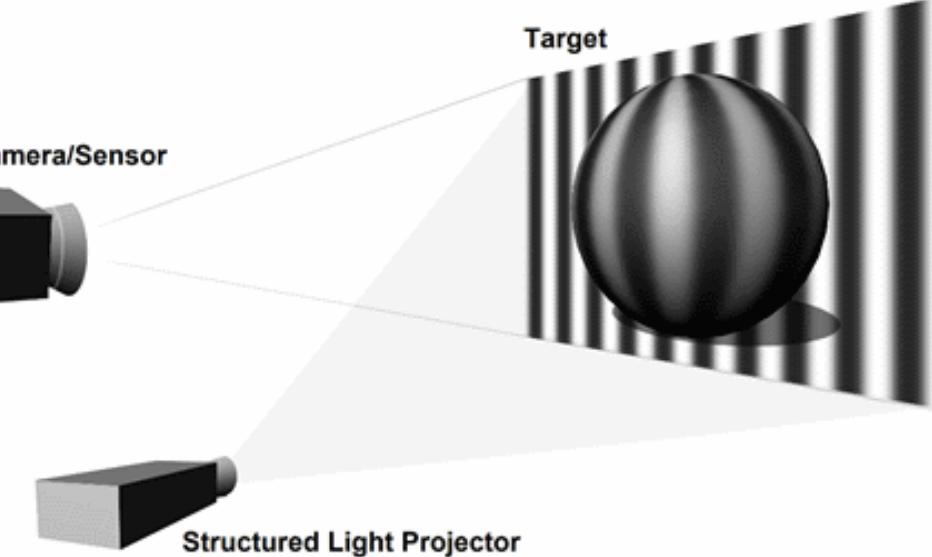
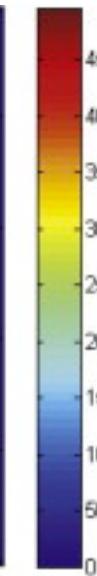
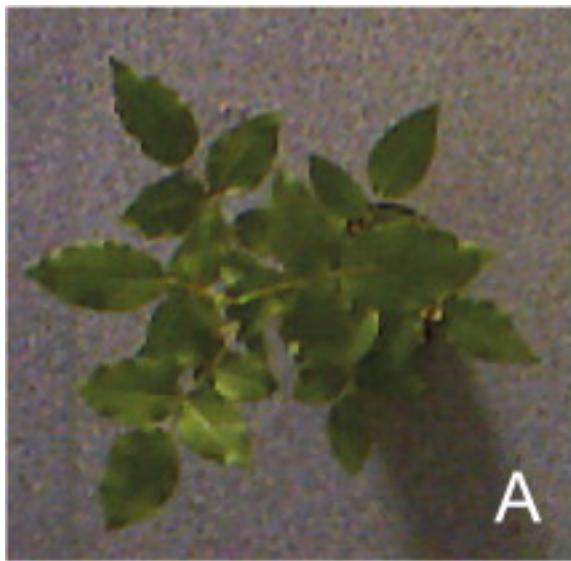
(Artzet et al., under revision)

Reconstructing architecture of small plants

## Contactless method



- Structured lighting



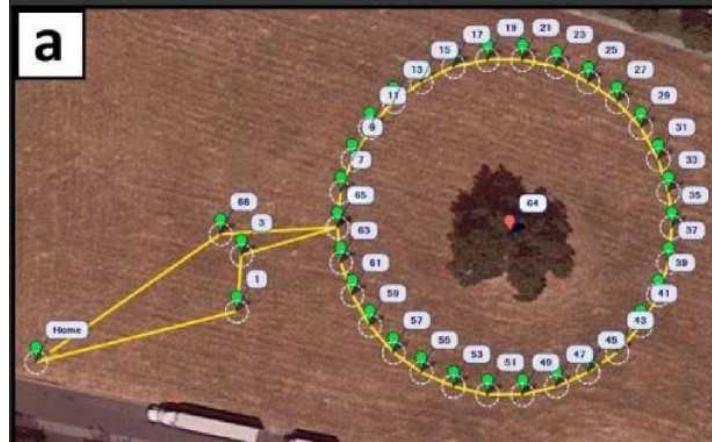
Chené et al., 2012.

Reconstructing architecture of small plants

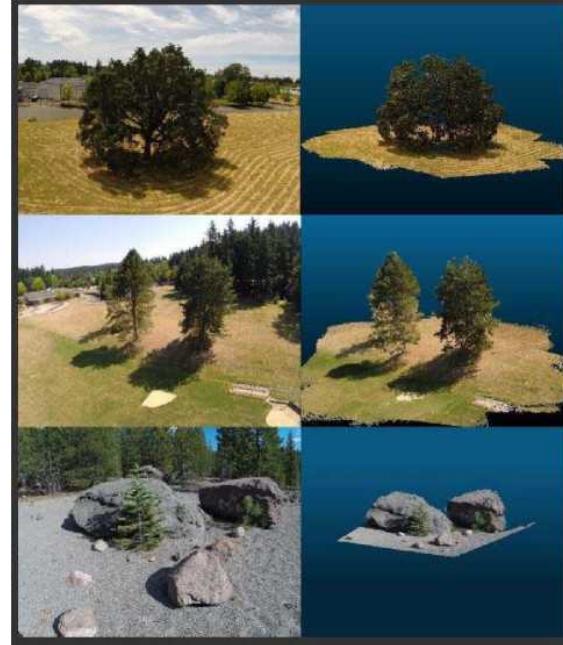
## Contactless method



- Photogrammetry - 3D by triangulation



Gatziolis D, Lienard JF, Vogs A, Strigul NS (2015) 3D Tree Dimensionality Assessment Using Photogrammetry and Small Unmanned Aerial Vehicles. PLoS ONE 10(9)



Precision depends on the number of views.  
Need to tune the number of view on which a point  
should appeared.  
Result in a point cloud.  
Could allow estimating LAI, LAD, Volume, Height.

## Contactless method



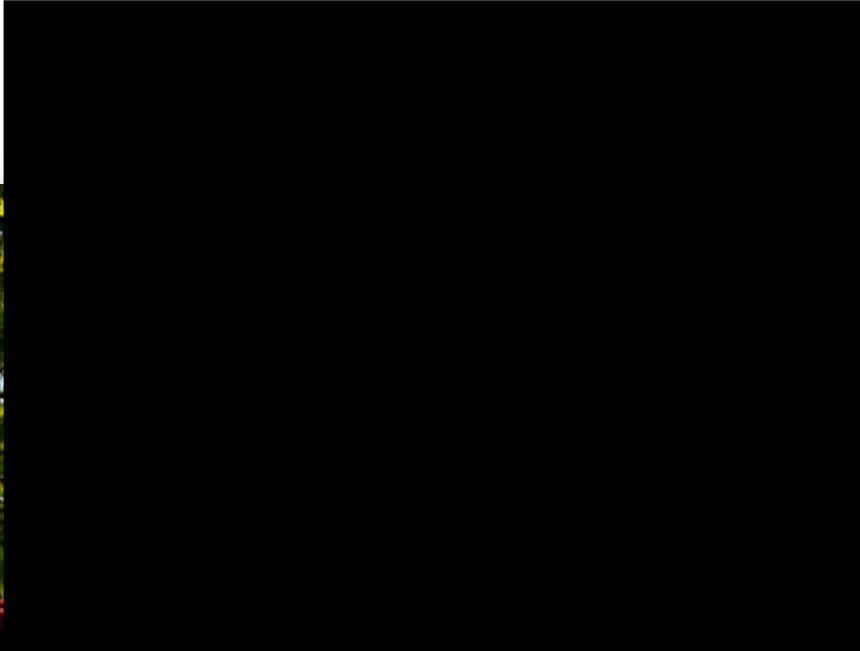
- Photogrammetry - 3D by triangulation



# Scan from terrestrial laser



- Laser scanner devices acquire 3D images of a plant



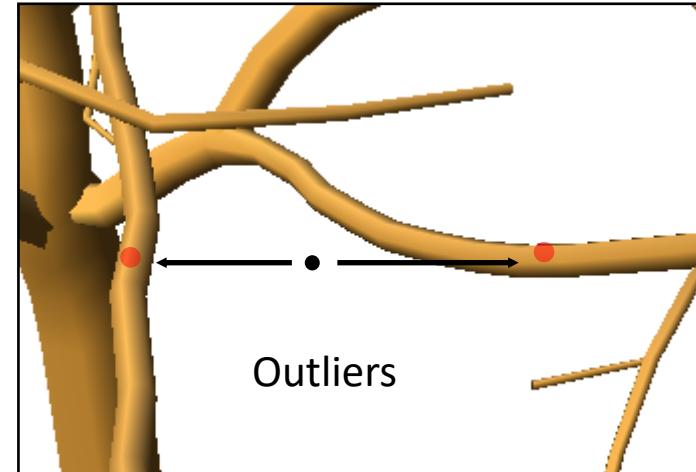
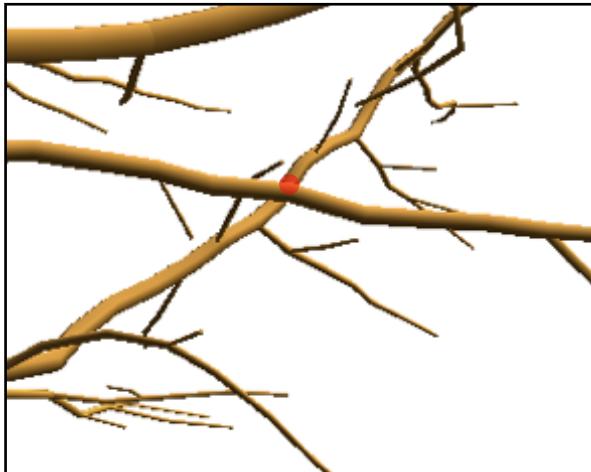
# Scan of plants



- Unstructured point set



- Several sources of noise :
  - Distance , Environmental condition (wind), Laser signal quantification, Laser footprint.



# Point Patterns



Dense data points  
(trunk, big branches)

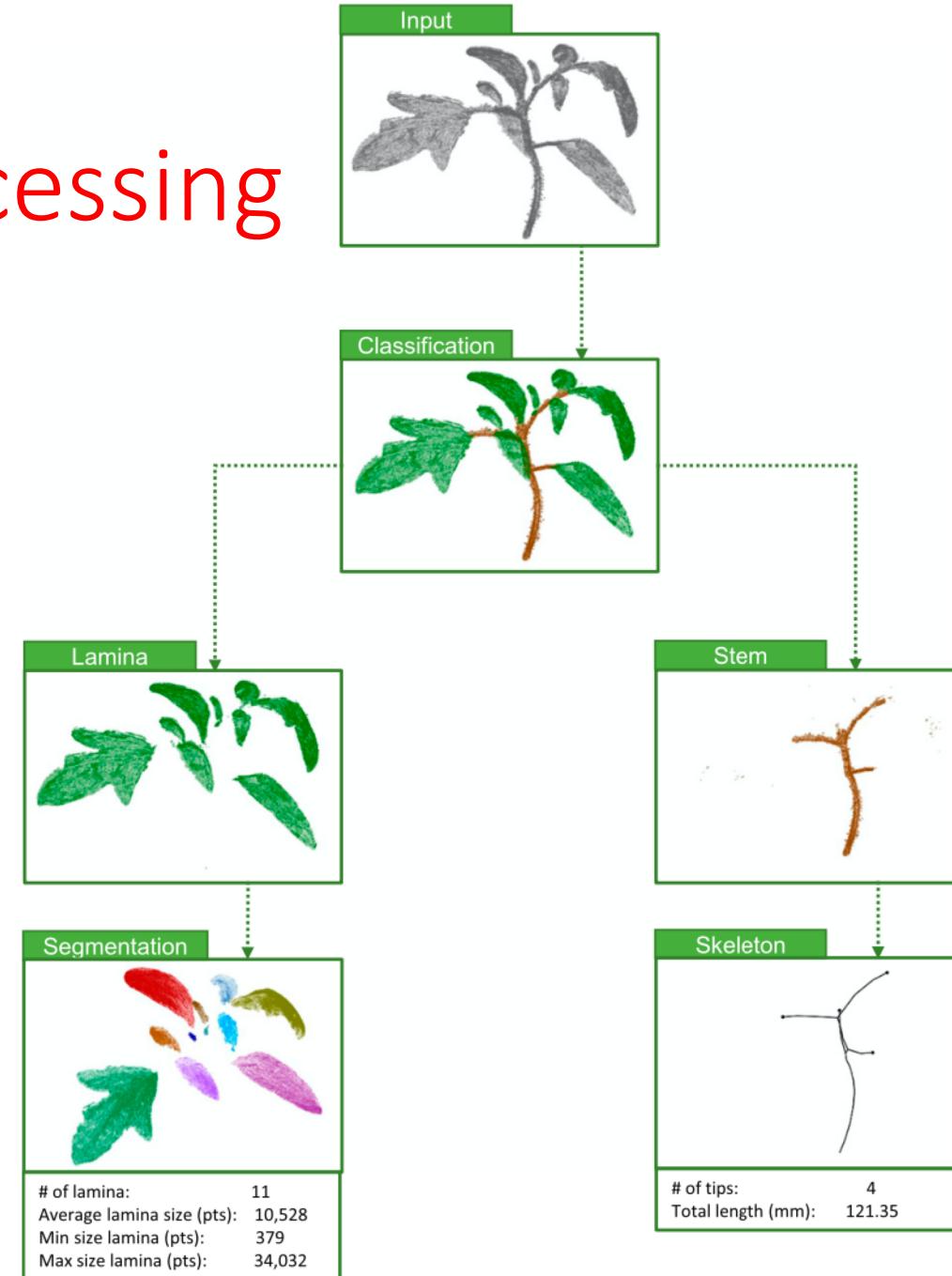
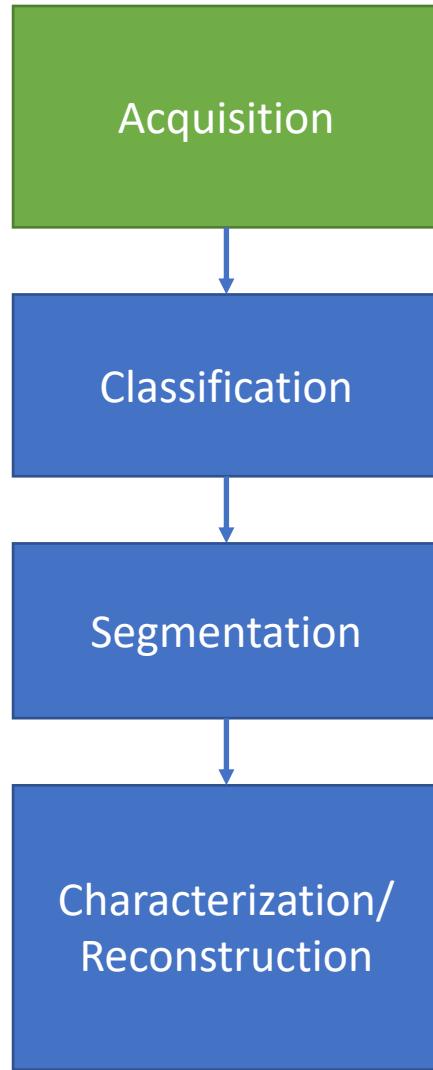


Sparse data points  
(small branches, twigs)



Incomplete data points  
(occlusion)

# Point cloud processing



Input

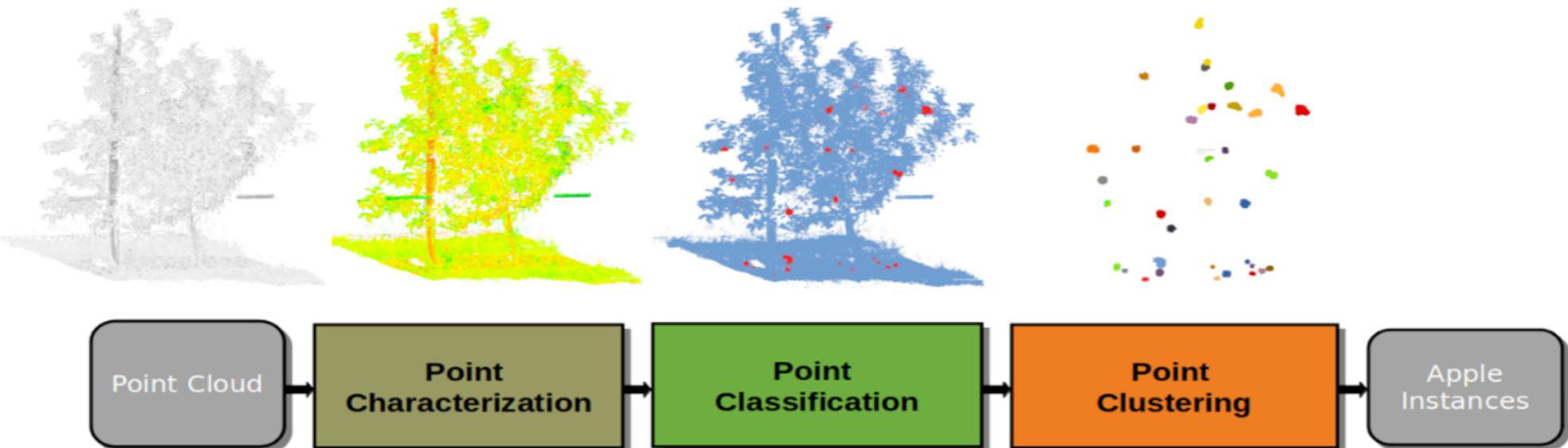


Classification



# Classification and segmentation: Machine learning based pipelines

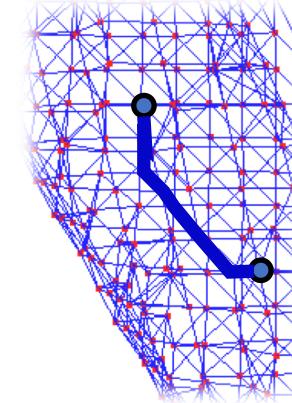
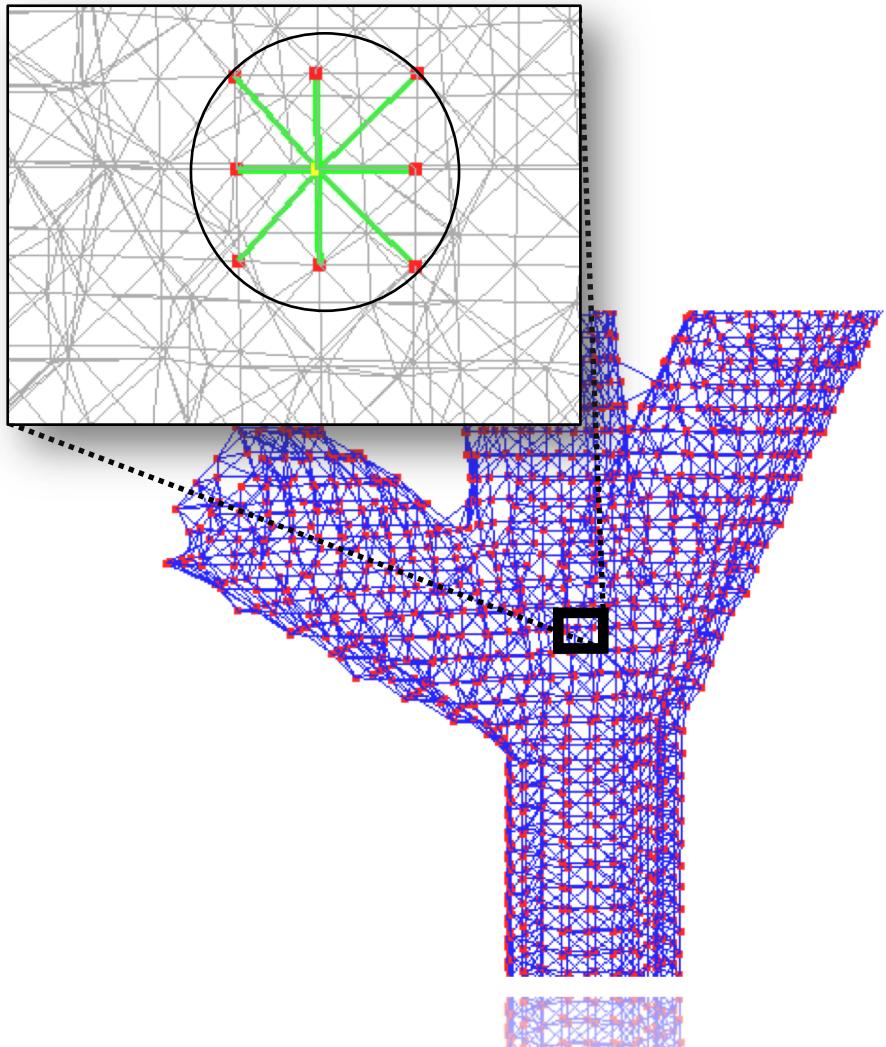
- Application on fruit detection (Artzet et al., 2020)



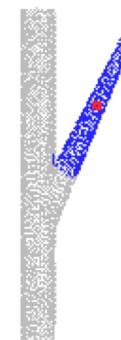
# Points Local Connection



- Each point is connected to its  $k$  closest neighbors. This form the Riemannian graph  $L$ .
- Distance between 2 points can be redefined as length of shortest path between them.



- The  $R$ -neighborhood of  $p_i$  is defined as points  $p_j$  whose distance of path to  $p_i$  on the graph  $L$  is smaller than  $R$ .



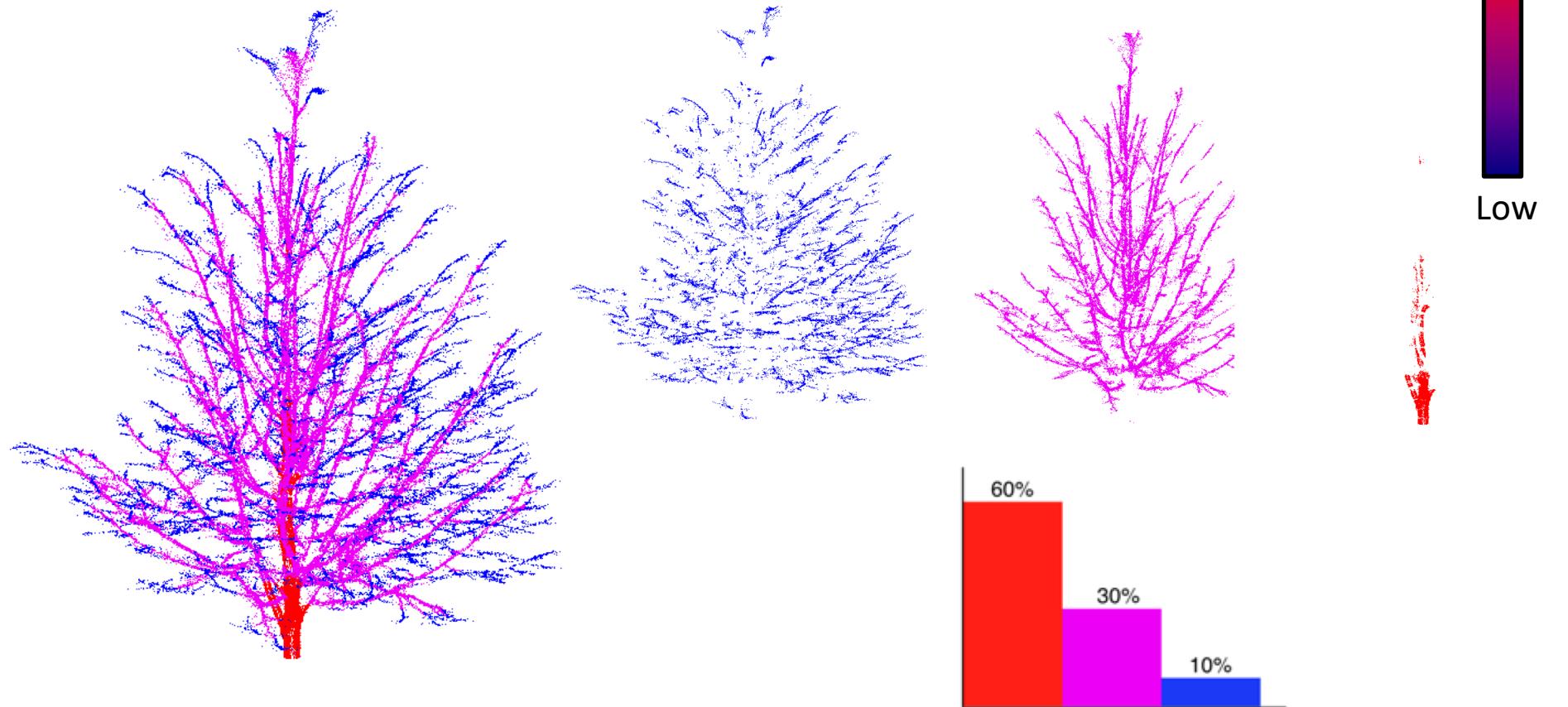
Distance on graph  $L$   
 $R=20$



# Points Local Density

- The local density depends on the number of points in the r-neighborhood. (Pauly et al, 02)

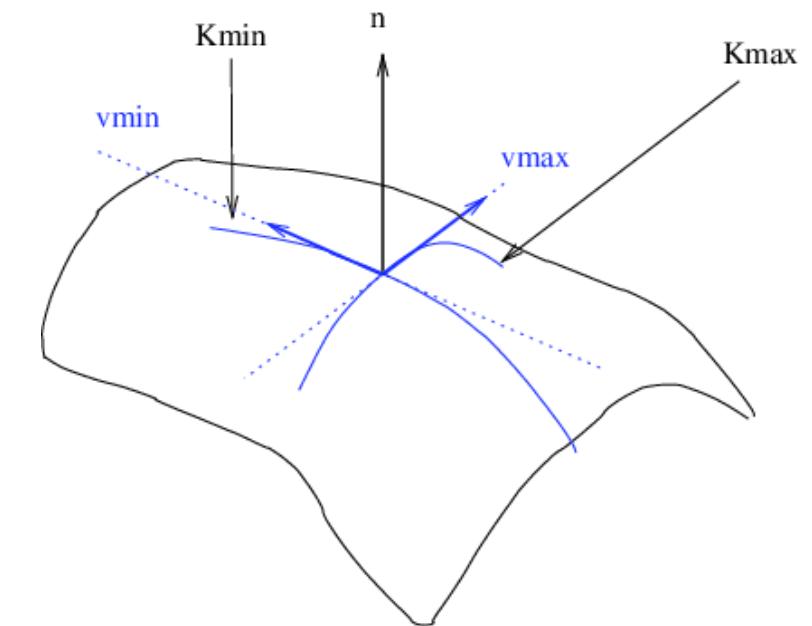
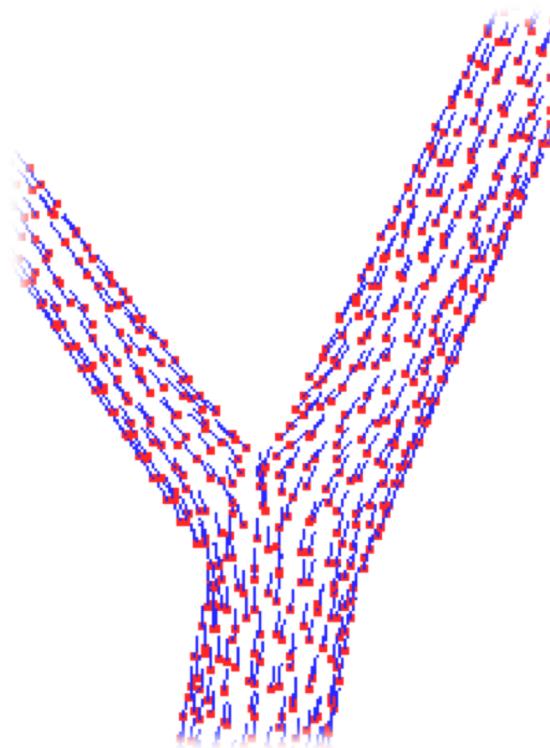
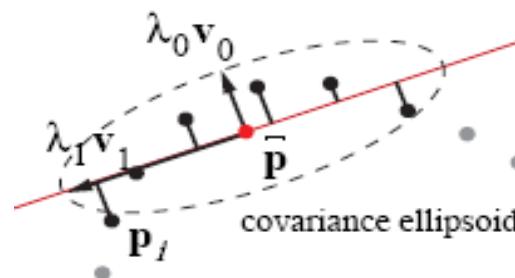
$$\lambda = \frac{k}{r^2}$$



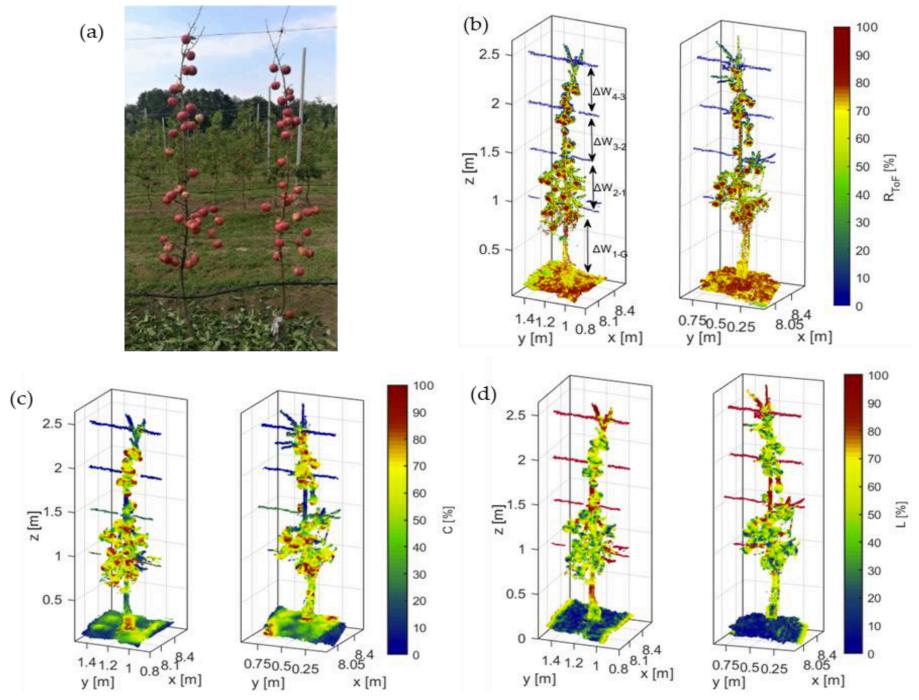


# Points Local Orientation and Curvature

- For each point, a Principal Component Analysis is made on its R-neighborhood. Principal direction gives local orientation of the point set.

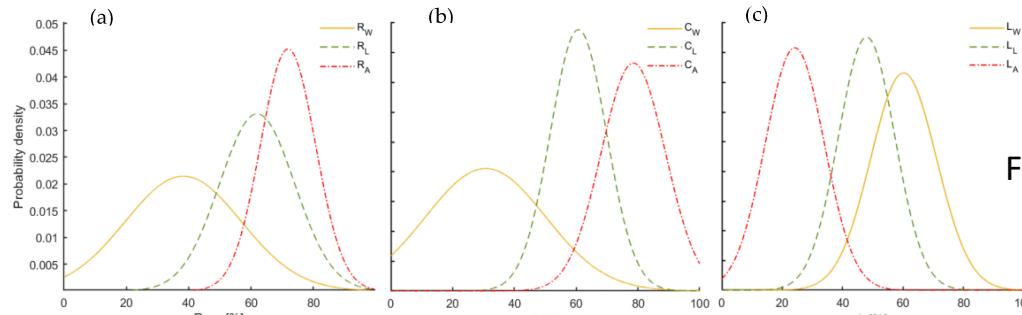


# Using spatial and radiometric features for segmentation



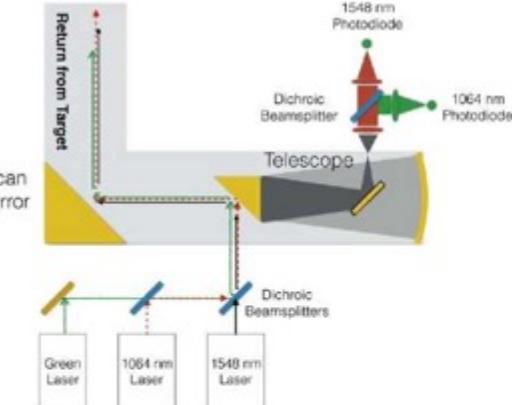
**Figure 4.** Representation of 3D point clouds of (a) RGB image (b) reflectance ( $R_{\text{ToF}}$ ) [%], (c) curvature (C) [%] and (d) linearity (L) [%] in defoliated trees at DAFB<sub>120</sub>.

From Tsoulias et al., 2020



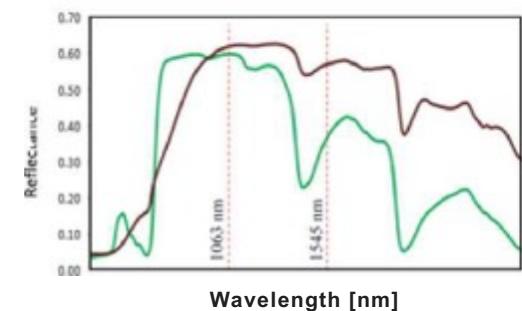
**Figure 5.** The probability density of (a) calibrated reflectance intensity ( $R_{\text{ToF}}$ ) [%], (b) curvature (C) [%], and (c) linearity (L) [%] for wood, leaves, and apples.

Toward multispectral LiDAR  
(foliage vs wood)

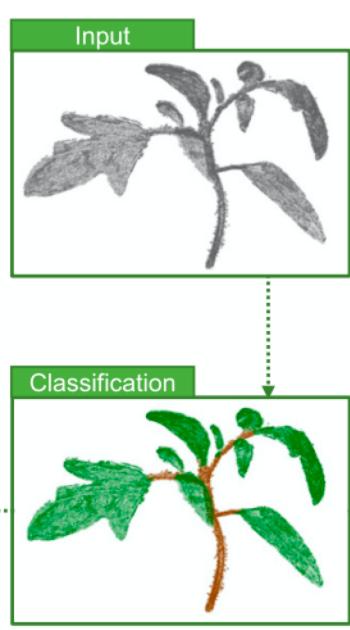


**Fig. 1.** Instrument schematic. Thin lines represent the outgoing coaxial beams. Gray lines represent the FOV of the instrument and the path of the observed returns. The continuous wave green laser serves as a marker and is not recorded.

From Douglas et al., 2015

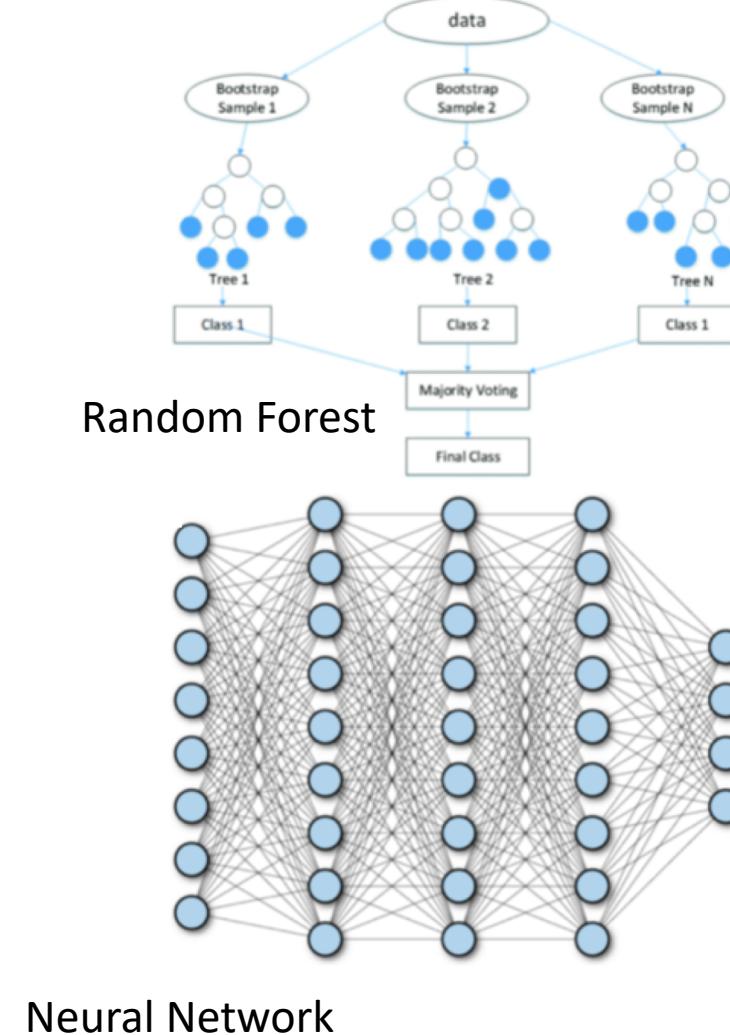
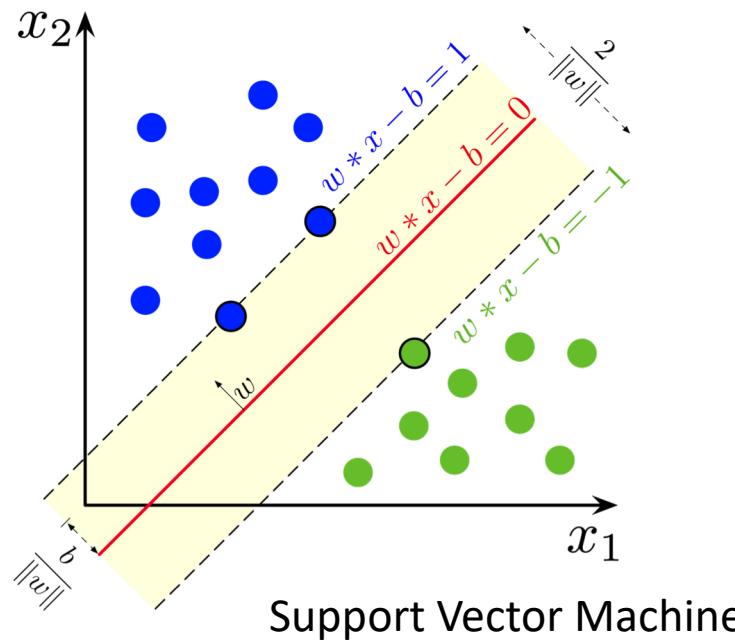


**Figure 1.** Spectral reflectance of an oak leaf (green) and oak bark (brown). From Danson et al., 2018



# Classification: Machine learning based pipelines

- Classification points of plant scan based on machine learning (Ziamtsov et al., 2019, Artzet et al., 2020)



Input

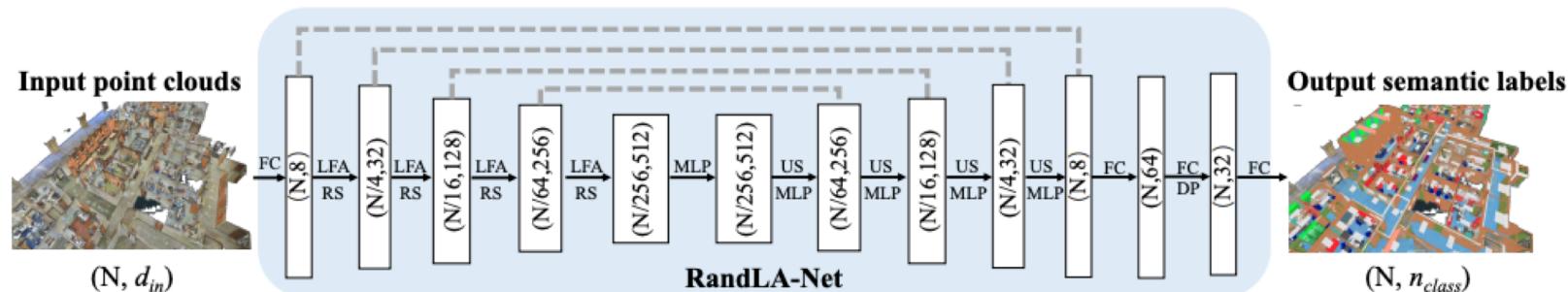
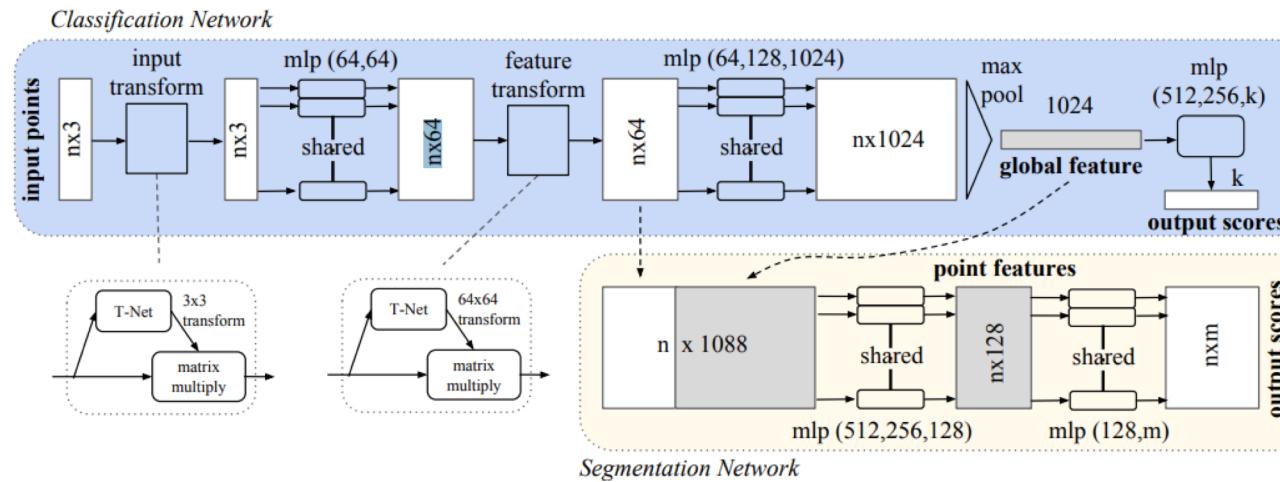


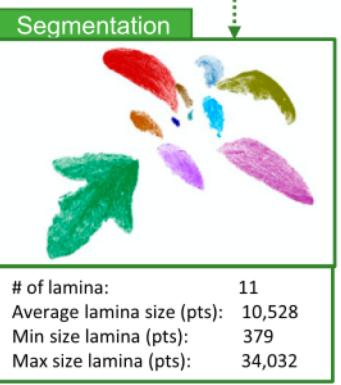
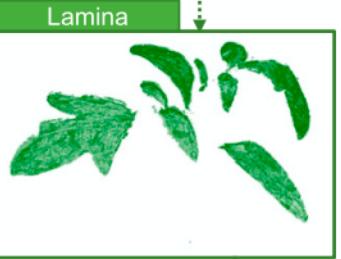
Classification



# Classification: deep learning based pipelines

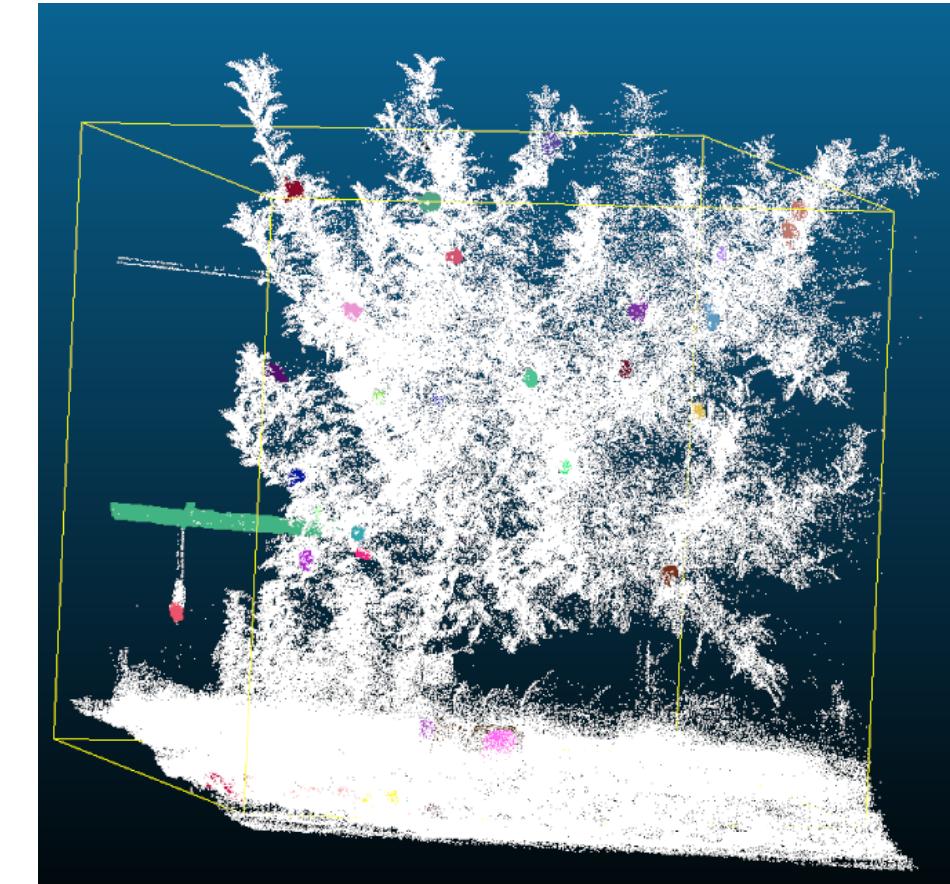
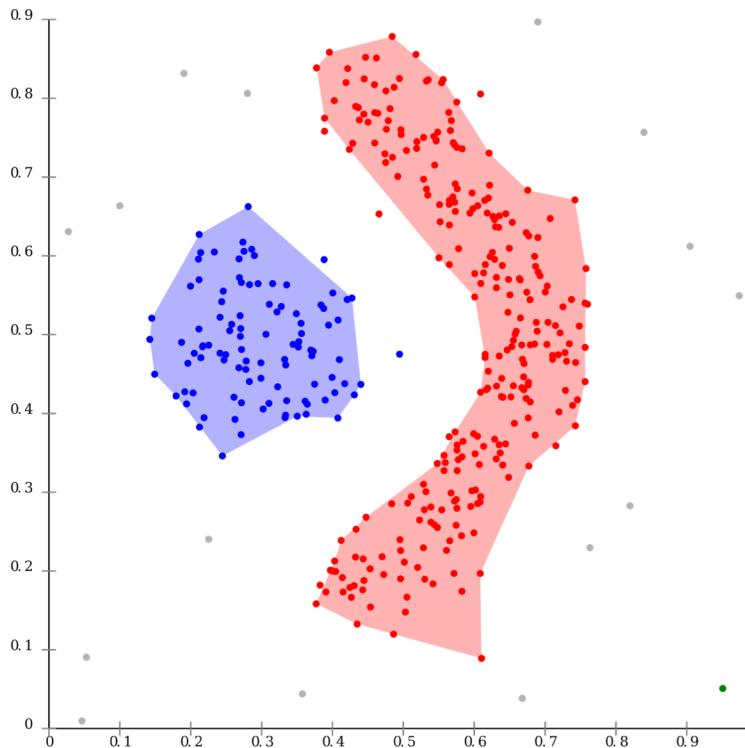
- Classification based on deep learning such as PointNet (Qi et al., 2016) or RandLaNet (Hu et al., 2019). Tested on plant in (Artzet et al., 2020)

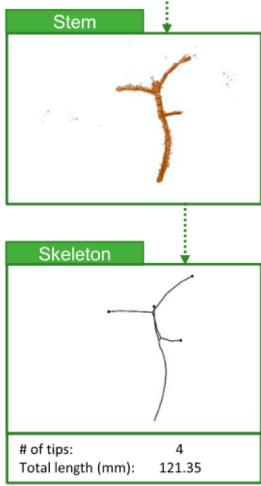




# Segmentation: clustering

- DBScan algorithm
  - Growing regions
  - Non linear





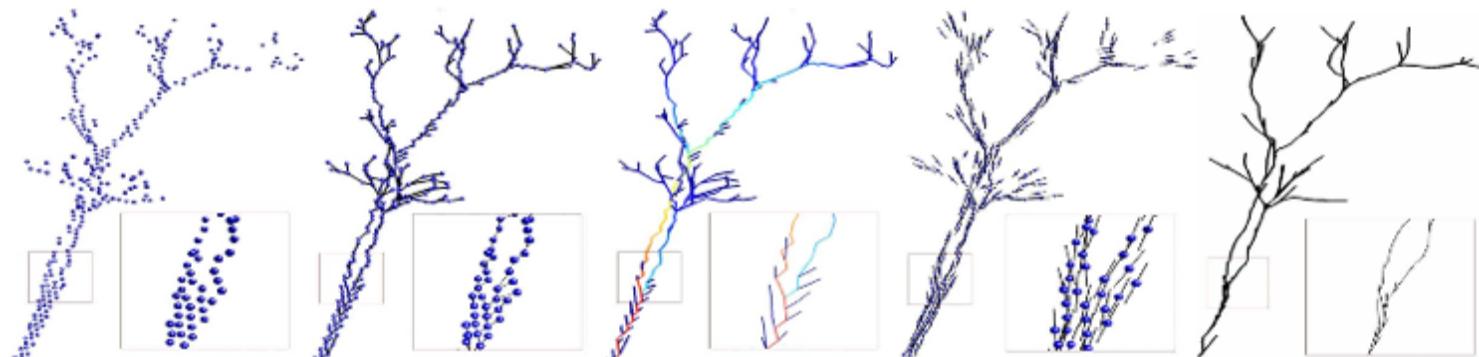
# Reconstructing plant architecture based on 3D-points



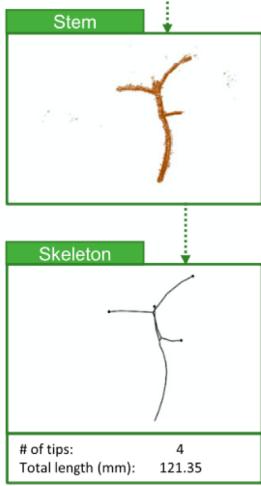
- Clustering of point set (Xu *et al.*, 07).



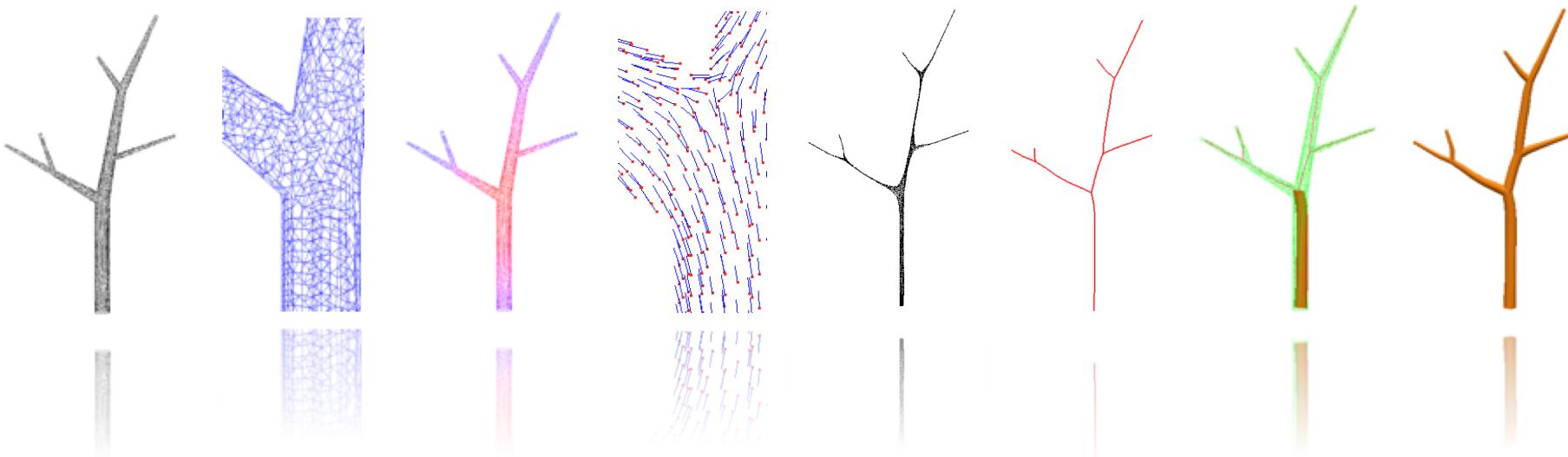
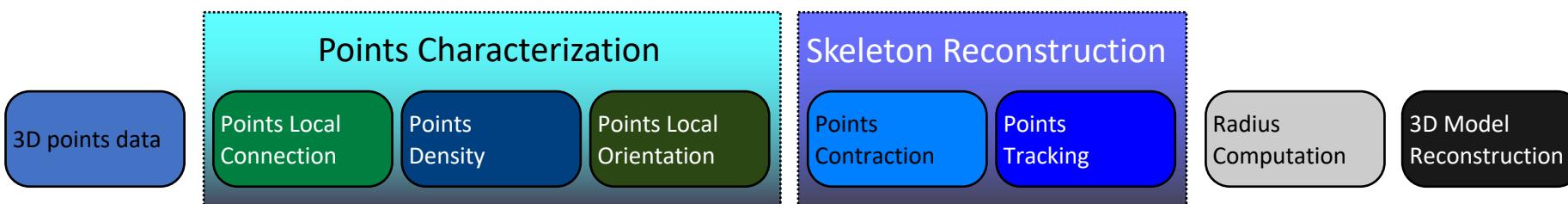
- Global optimization and point alignment to reconstruct branching systems (Livny *et al.*, 10).



(see TP1)



# A Pipeline for Reconstruction 3D Architecture (Preuksakarn et al., 2013)



(see TP1)

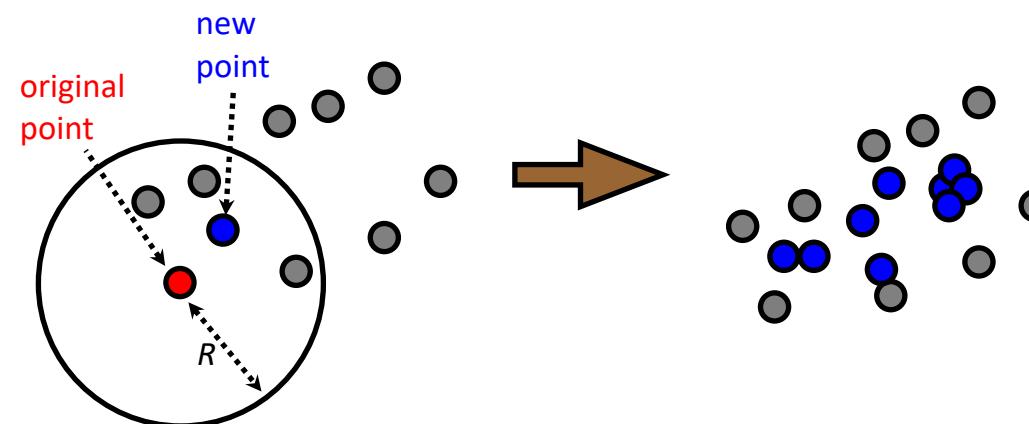


# Points Contraction

- Shift the points from the border of the shape toward its center (Riccardo et al, 99).

$$c_i^R = \frac{\sum_{p_j \in S_i^R} p_j}{|S_i^R|}$$

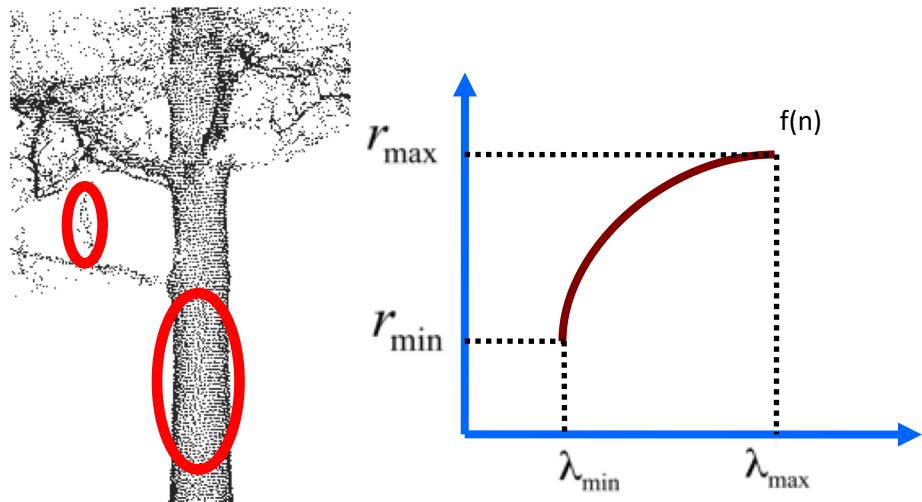
$$P \rightarrow P^R = \{c_i^R\}$$



# Adaptive Distance



- Distance of contraction will be determined according to the local density and user defined function.



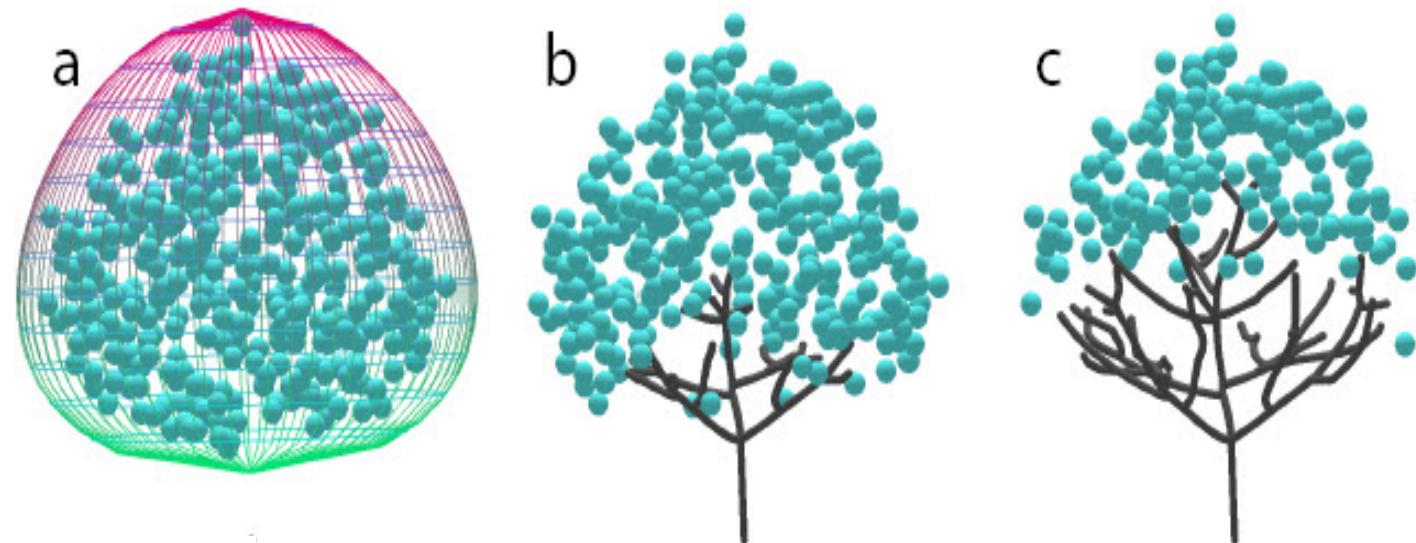
Use density information  
to discriminate





# Point Tracking Algorithm

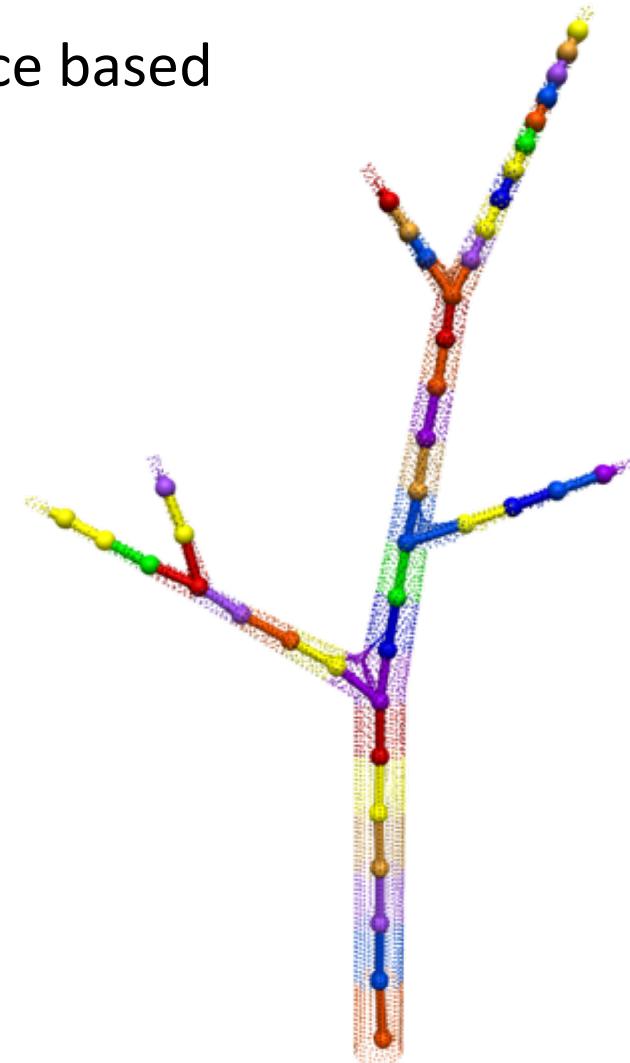
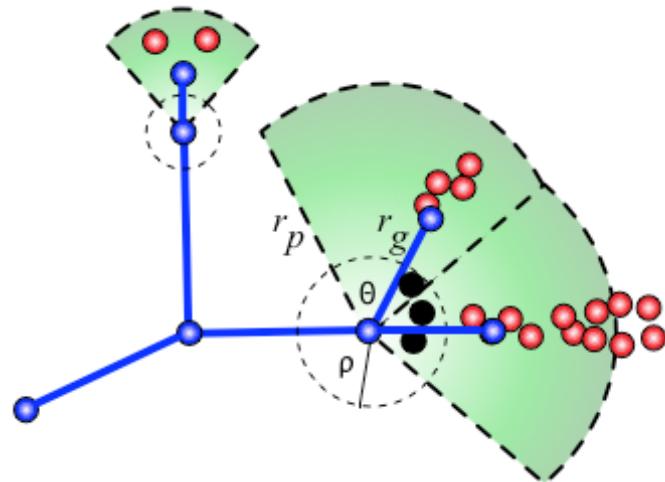
- Based on the Space Colonisation Algorithm (Runions et al., 07, Palubicki et al., 09). The branching structure of trees try to reach the attraction points.





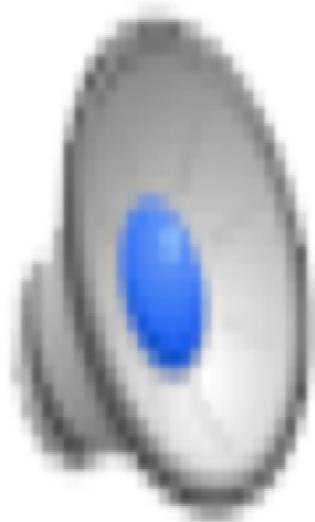
# Adaptive Point Tracking

- The growth, influence and kill distance based on local density.



- The colonization will be done in the Riemannian graph space.

# Reconstruction





# Estimation of the diameter

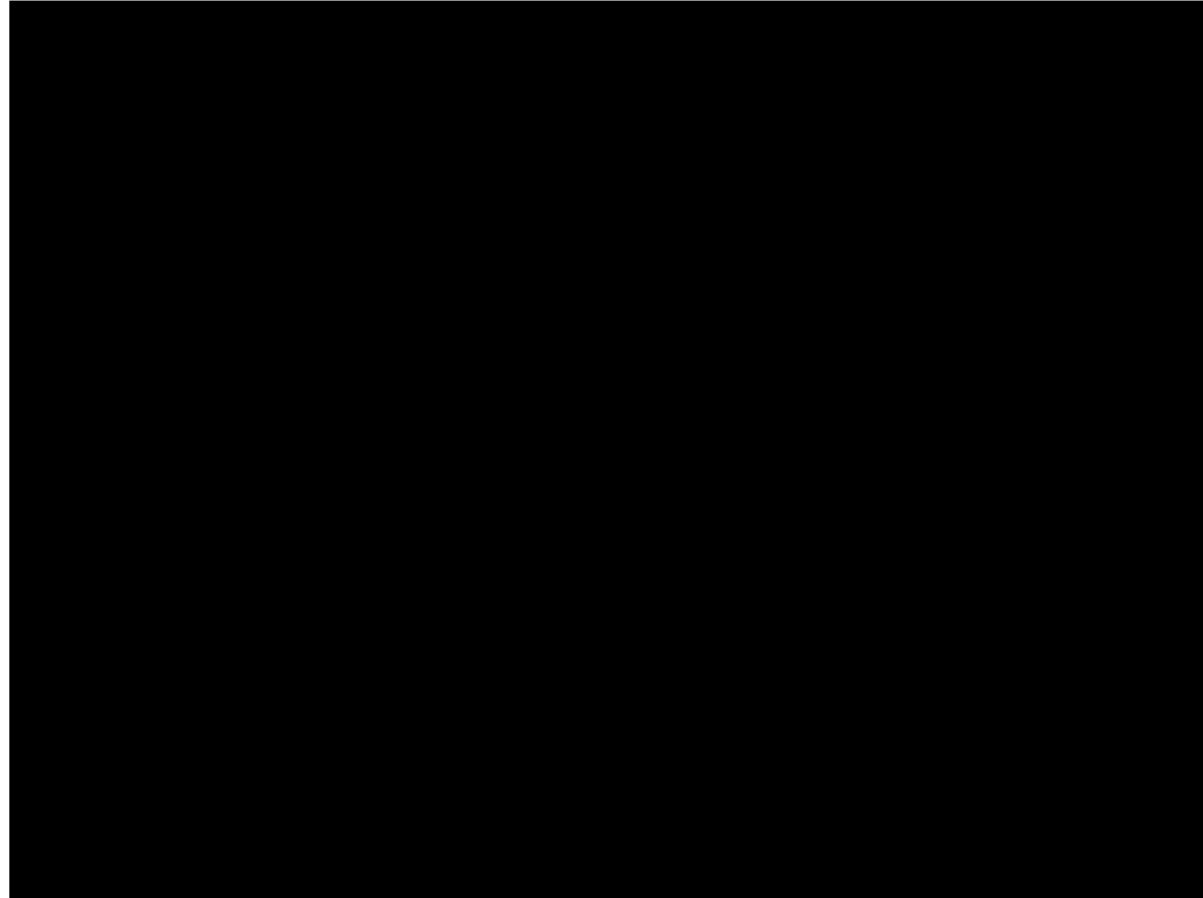
- The radius at each node is estimated using the pipe model [Shinozaki et al. 64].

$$d_i^p = d_j^p + d_k^p$$

$$p = \frac{\log |n_{f_i}|}{\log d_i - \log d_{f_i}}$$



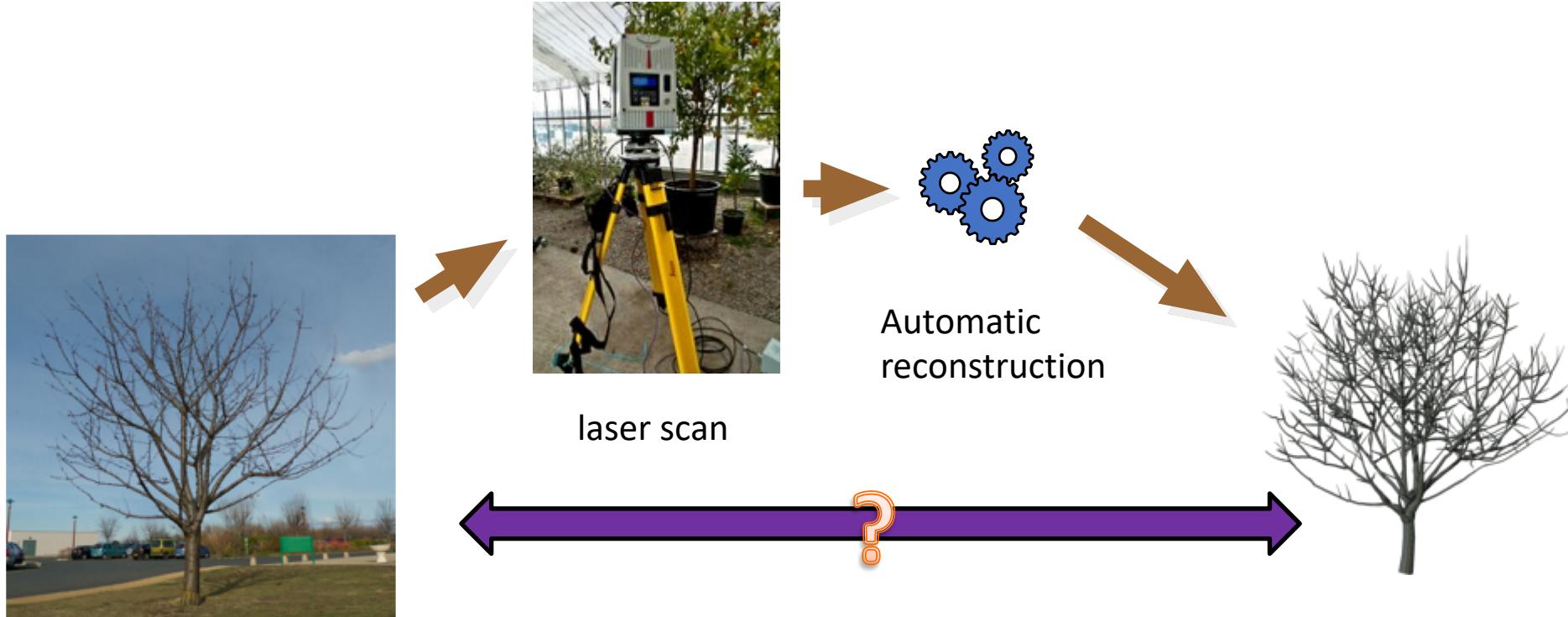
# Results



# Results



# Problem of the quantitative assessment of automatic reconstruction



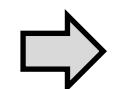


# Global Comparison

- Indirect index comparison (Côté *et al.*, 09, etc.)
  - ❖ Wood content, foliage area, crown volume, etc..
  - ❖ Reflectance signature, transmission properties, etc..



(Pradal, Boudon et al, 09)

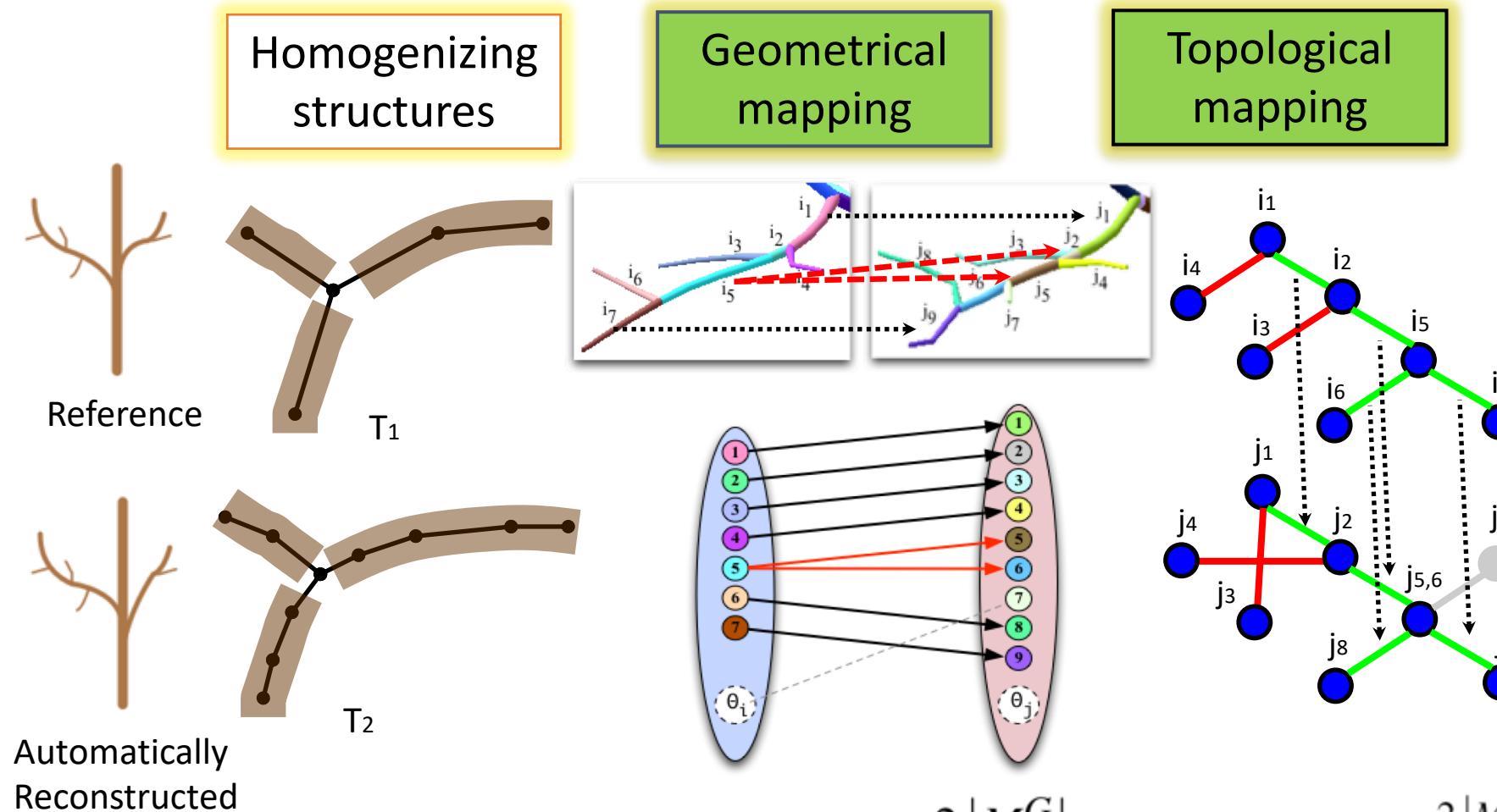


Evaluate indirectly the accuracy of the structure

# Structure Comparison



- Finding geometrical and structural similarities



$$D_G(T_1, T_2) = \frac{2 |M^G|}{|T_1| + |T_2|}$$

$$D_T(T_1, T_2) = \frac{2 |M^T|}{|E_1| + |E_2|}$$

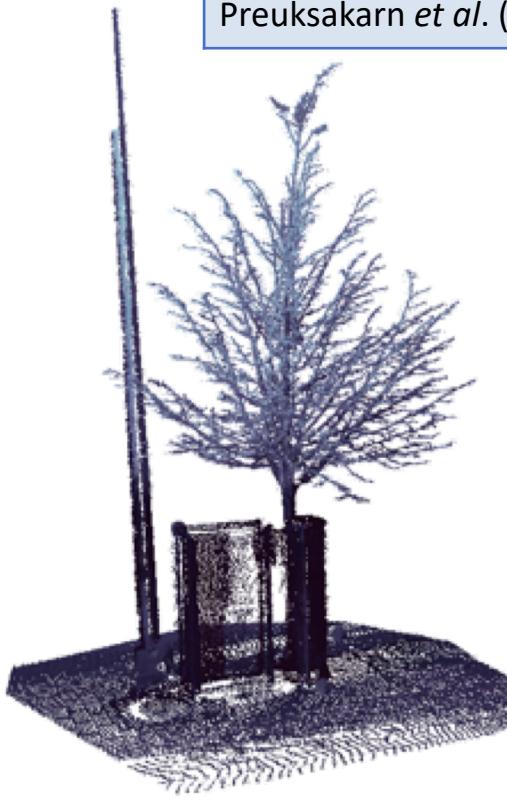
(Boudon et al., 2012)

# Comparison of methods of the literature



- Comparative evaluation of (Xu *et al.*, 07), (Livny *et al.*, 10) and (Preuksakarn *et al.*, 10).

Method	Mean Point Distance (mm)	Volume (Ref: 13.17m <sup>2</sup> )	$D_G$	$D_T$
Xu <i>et al.</i> (2007)	12.55	12.76	0.81	0.87
Livny <i>et al.</i> (2010)	6.49	12.7	0.69	0.65
Preuksakarn <i>et al.</i> (2010)	7.26	13.3	0.90	0.72



Xu *et al.*, 07



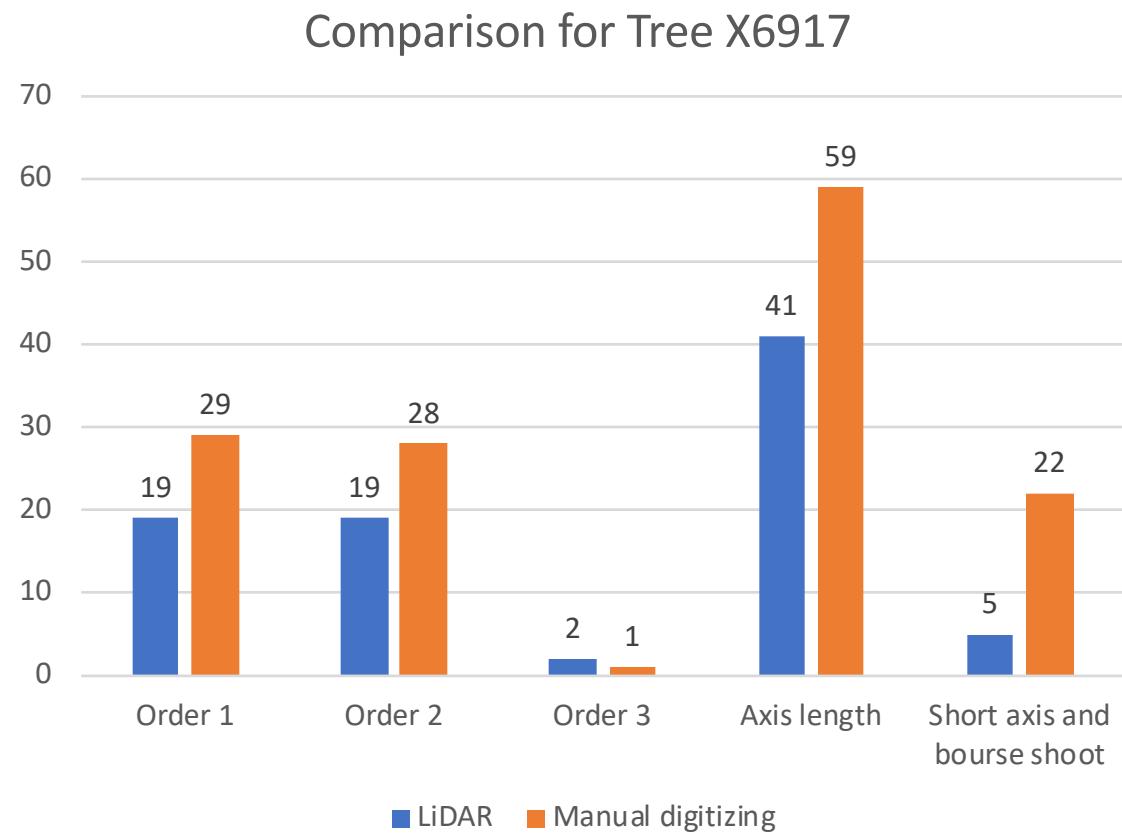
Livny *et al.*, 10



Preuksakarn *et al.*, 10



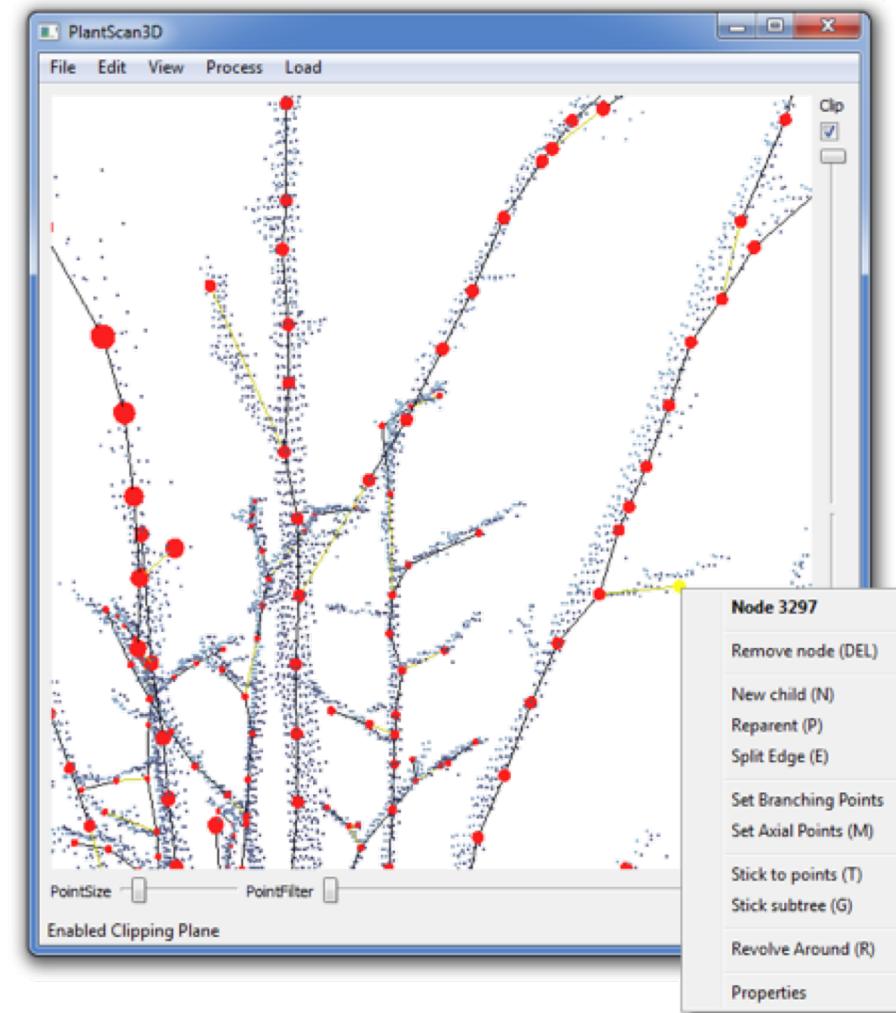
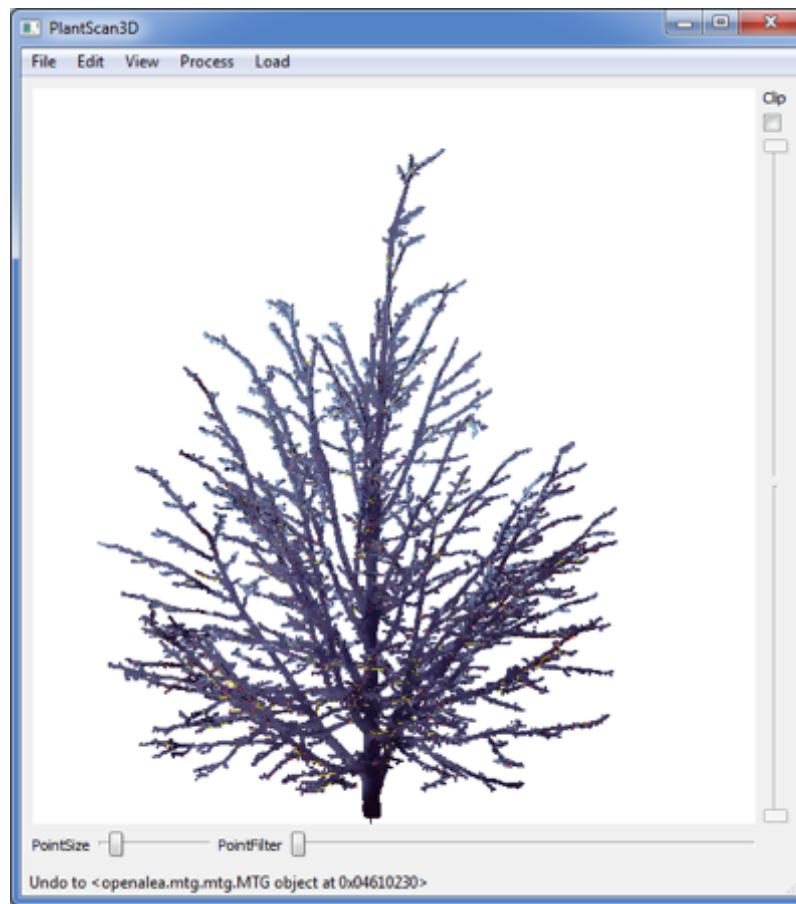
# Comparing reconstructions from LiDAR and magnetic digitizing



Differences come mainly from short branches and bourse shoot.



# Editor of plant architecture from laser scans



See TP

# Foliage reconstruction based on allometric relationship

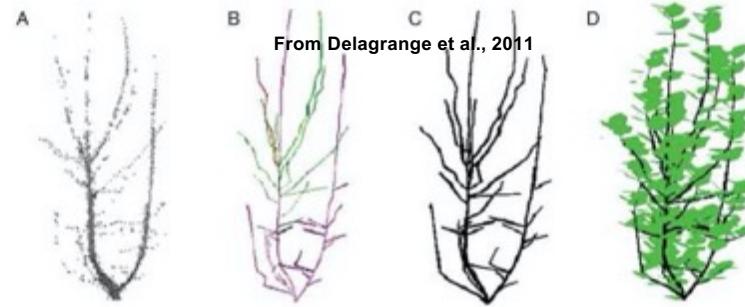
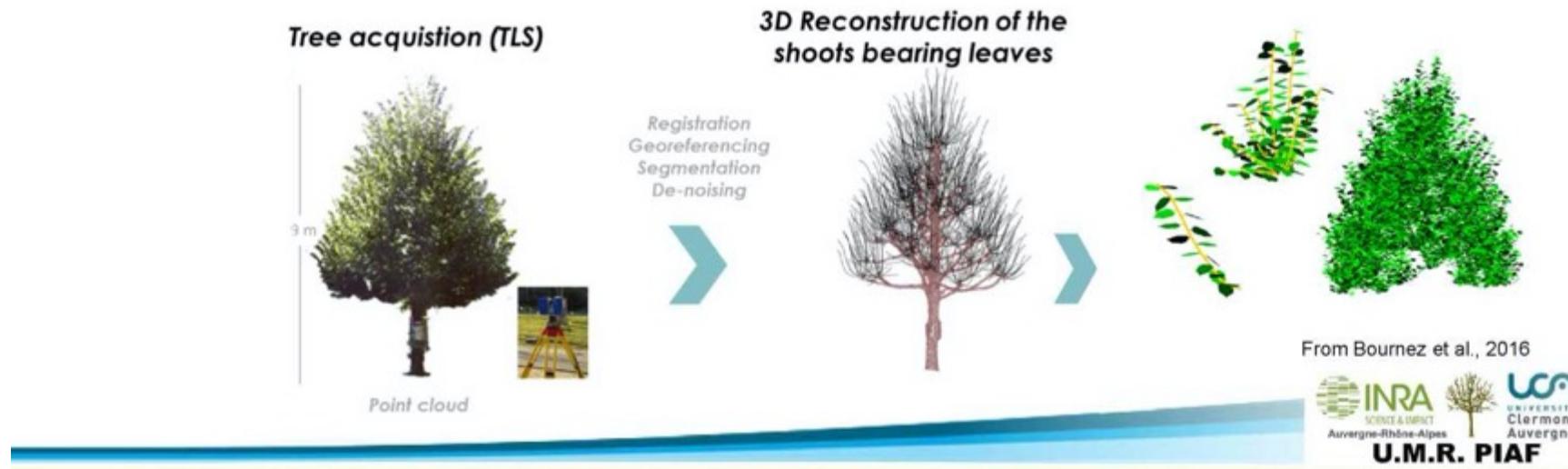
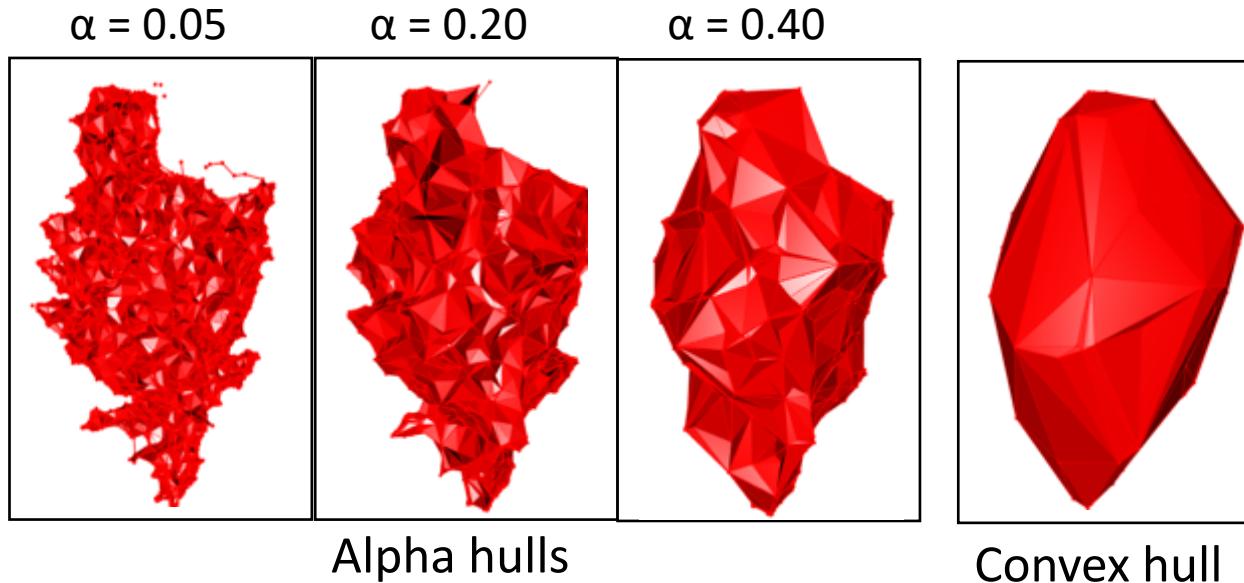


FIG. 2. Main steps of crown reconstruction from T-LiDAR scans. (A) Registered point cloud of the leaf-off scene (B), reconstructed skeleton, (C) pipe representation with reconstructed radius (D) and positioned leaves on the reconstructed pipe representation.

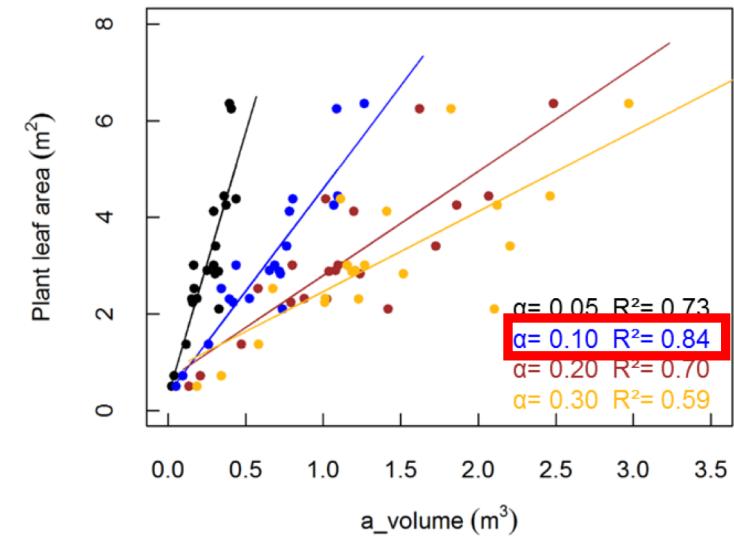


# Foliage estimation based on allometric relationship

- Estimation of alpha volume ( $a\_volume$ )



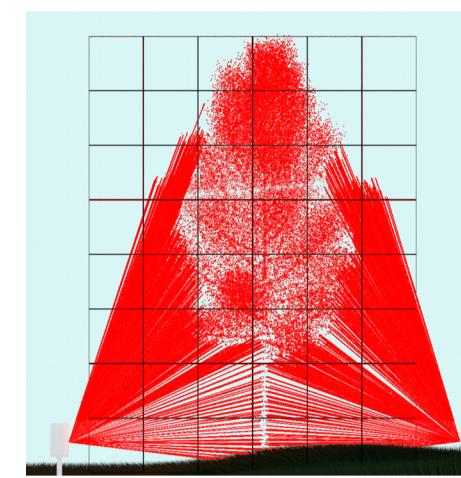
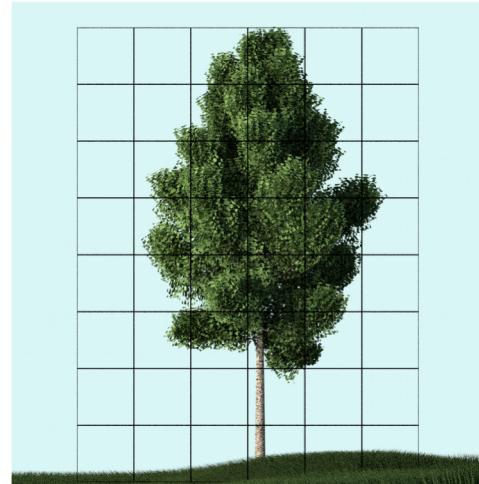
- Relation with foliage area



(Pallas et al., 2018)

# Voxelization based foliage reconstruction from Lidar

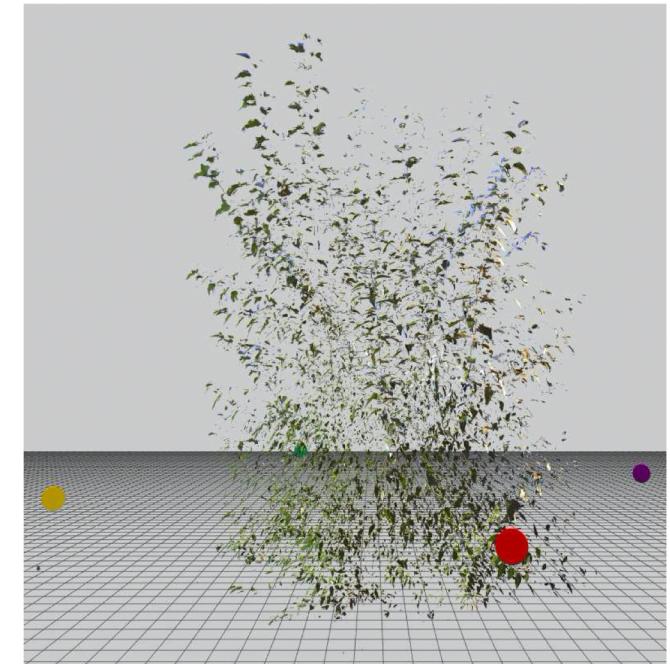
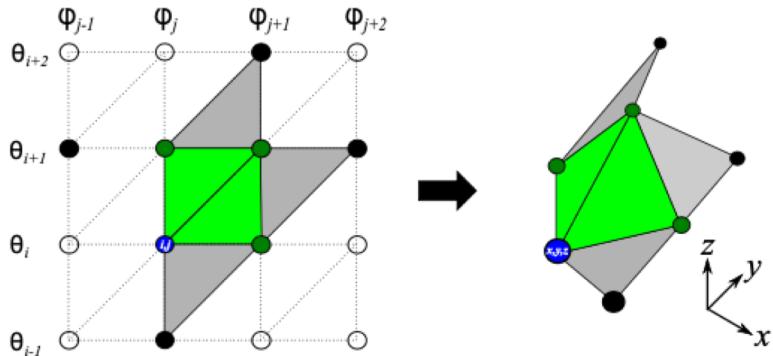
- Estimation of leaf area from transmittance of voxels from input and output lidar rays in each voxels



(Gregoire et al., 2017)

# Foliage triangulation from Lidar

- Filtered triangulation of foliage (Bailey et al., 2017)



# Conclusion

- A rapidly improving field.
  - New acquisition devices
  - New processing techniques (deep learning).
- Testing different protocols and devices

	Lidar	Digitizing	Photogrammetry
Cost	>30 k€	~ 8k€	~ 8k€
Range	~100 m	~ m	~ 100 m
Precision	++	+++	+
Software	In development	Operational	In Development /Operational

# Training

[https://github.com/openalea-training/hbma312\\_training/blob/master/laserreconstruction/Reconstruction%20from%20laser%20scans.ipynb](https://github.com/openalea-training/hbma312_training/blob/master/laserreconstruction/Reconstruction%20from%20laser%20scans.ipynb)

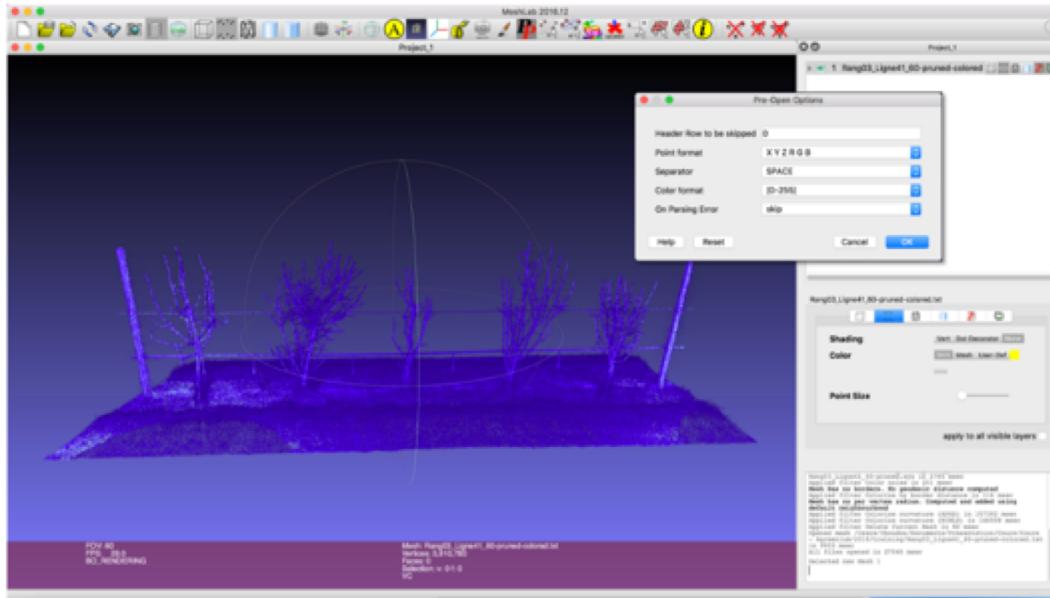
jupyter Reconstruction from laser scans Last Checkpoint: 16 hours ago (autosaved) Logout

File Edit View Insert Cell Kernel Help

Not Trusted Python 2

Importing the point cloud

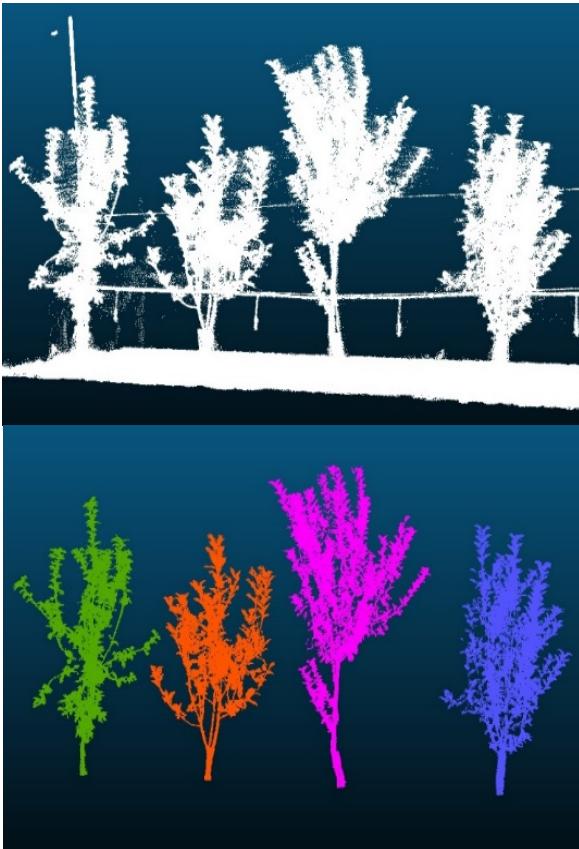
Let's consider the point cloud corresponding to the co-registration of several scans that you can download [here](#) (unzip it before using it). It contains several trees and environmental objects such as soil, wires and stakes. To have a first feeling of the scan, you can open it with [MeshLab](#). You can import the point set in MeshLab with the menu File>Import Mesh. A window will ask you to select parameters of import of the file. For this particular file, you should choose Point Format with X Y Z R G B and Separator with SPACE.



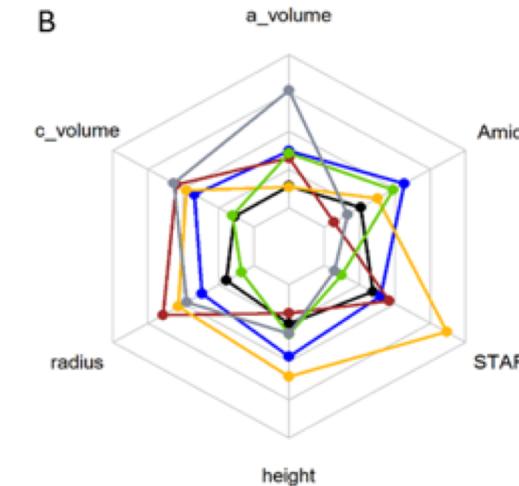
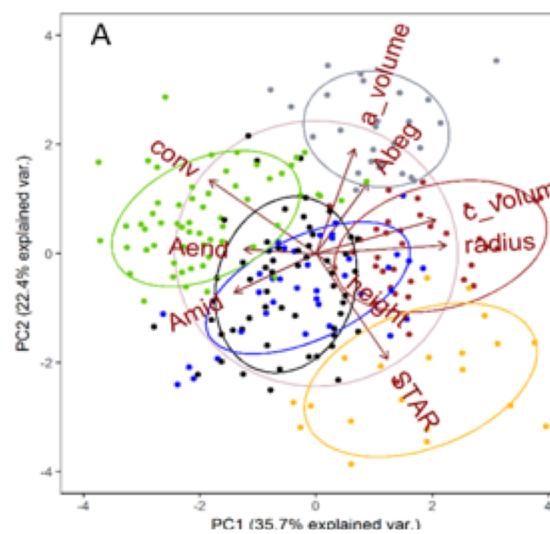
To have better rendering of the point cloud, you can play with display parameters on the right panel. In particular, for Shading use None. You can also control the Point size.

# Application for the classification of genotypes into morphotypes

- Scan of a core collection

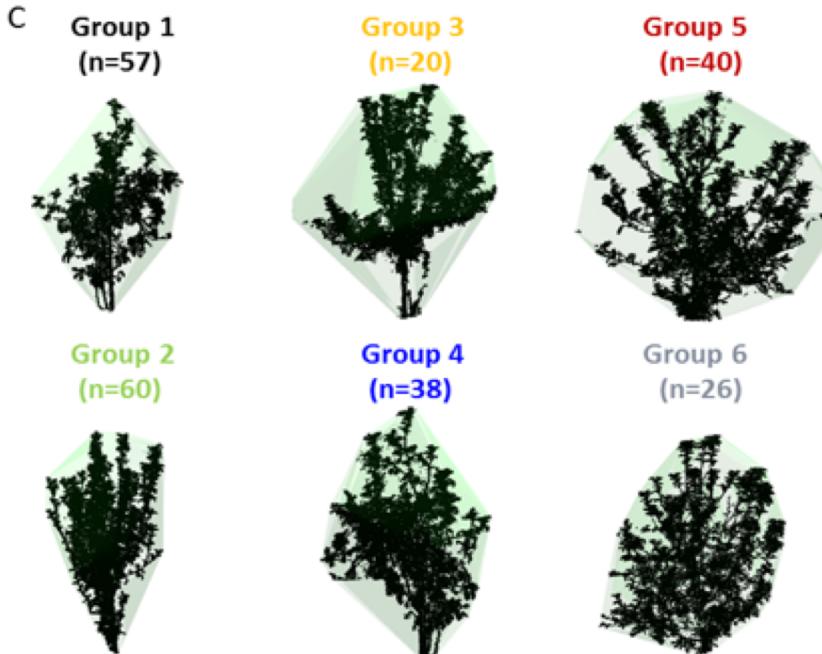


- PCA



- Clustering

- Group 1  
• Group 2  
• Group 3  
• Group 4  
• Group 5  
• Group 6



- Traits estimation

- Tree radius R and height H
- Convex volume (c\_volume)
- Alpha volume (a\_volume)
- Convexity (a\_volume/c\_volume)

radius, c\_volume

STAR, a\_volume  
Cf TD  
(Pallas et al., 2018)

Thank you