Image based modelling approaches for analysis of plant tissues morphology

<u>David Legland</u>, Thang Le, Anne-Laure Chateigner-Boutin, Christine Girousse



> General context:

valorization and transformation of agro-resources

- Applications
 - Food
 - Bio based materials
 - Biofuels







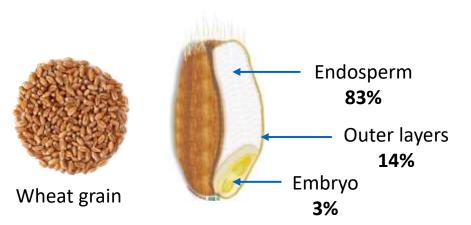






> Modelling growth of wheat grain - context

- Wheat: major crop resource worldwide
 - Yields for human and animal feeding
 - Impact of grain shape on milling process





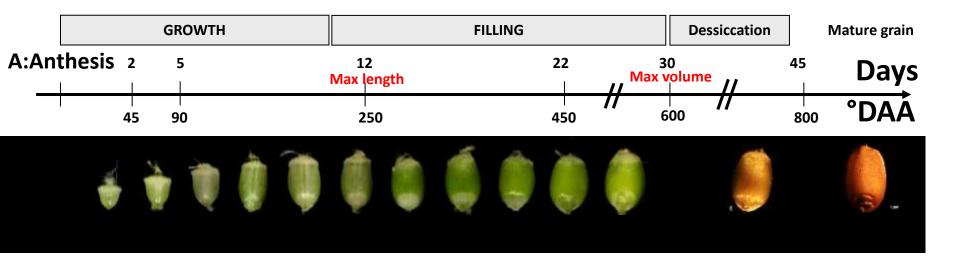


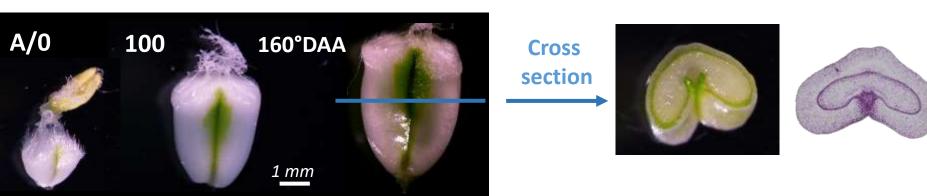
- Decrease on global yields + global warming
 - What are the processes that govern **the size and the shape** of the mature wheat grain?

INRAØ

> Objective: study of wheat grain growth

Changes of size and shape – towards morphogenesis



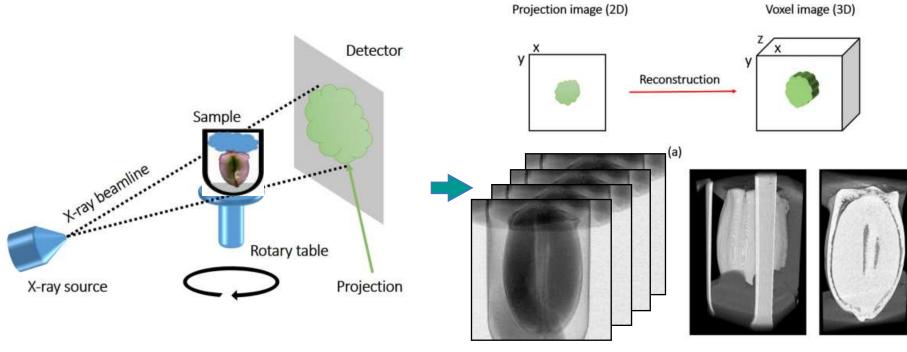


INRAØ

Image analysis and modelling for plant tissue morphology 2024-06-13 / AFH Saint-Malo / D. Legland

Need for whole-grain + 3D imaging

> 3D imaging using μ -tomography



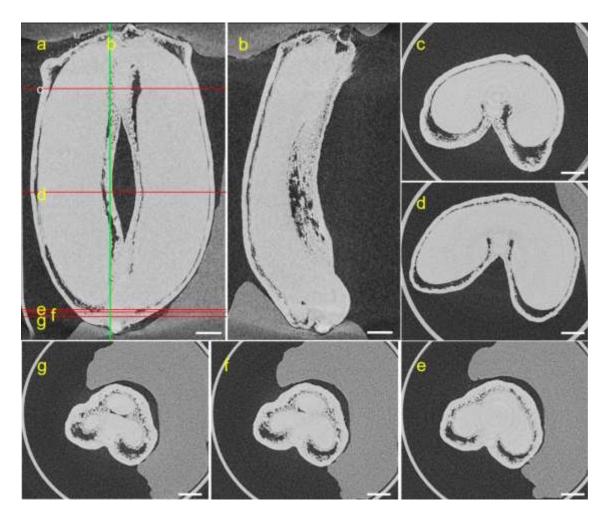
Nanotom 180 (G&E)

INRA

A stack of 2D CT slices



> 3D imaging by X-ray micro-tomography



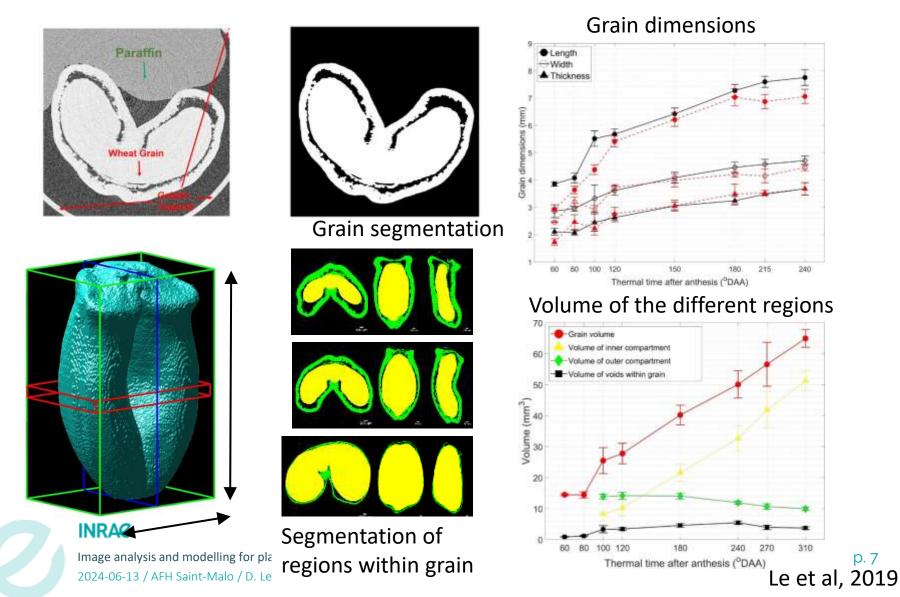
+ whole 3D imaging
+ tissue (and cells)
determination
+ fast (5-10 min)

- Destructive sampling
 - No time-lapse imaging...
 - Use series of static images



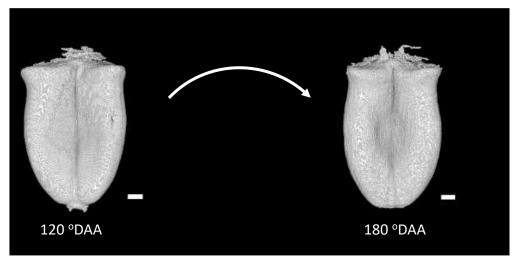
Study of growth by 3D image analysis

Quantification of global size & shape features for each stage





 How to better describe the changes of morphology between two successive stages?



Scale bar is 600 µm

 Seek for the geometric deformation between grains at two successive stages





Search for the optimal deformation by using shape registration:

$$\phi^* = \operatorname*{argmin}_{\phi} \mathcal{D}(\phi(S), R) + \mathcal{R}(\phi)$$

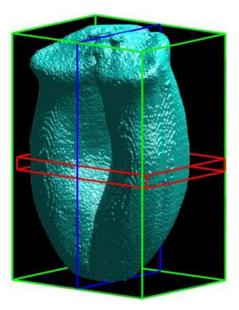
- *\phi*: deformation model
- S, R: individual ('subject') and reference shapes
- D: dissimilarity metric
- **R**: regularization function

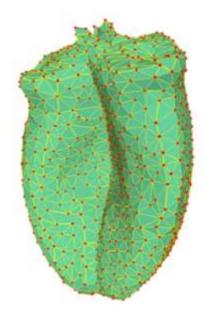


INRAe



- Transformation into 3D triangular meshes
 - (geometrization of images)
 - Reduction of computational complexity
 - Simplification of results interpretation



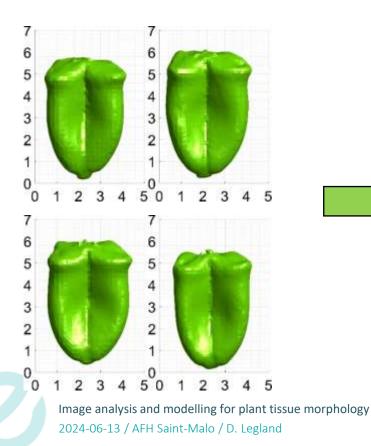


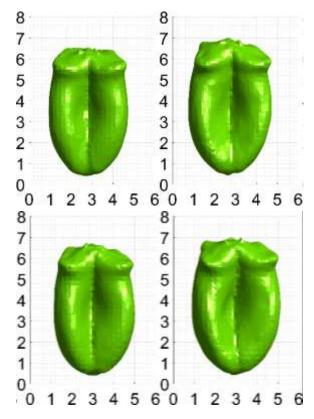


> Taking into account the biological variability

Group-wise registration

- Each stage is represented by several grains
- Need to register population of shapes...

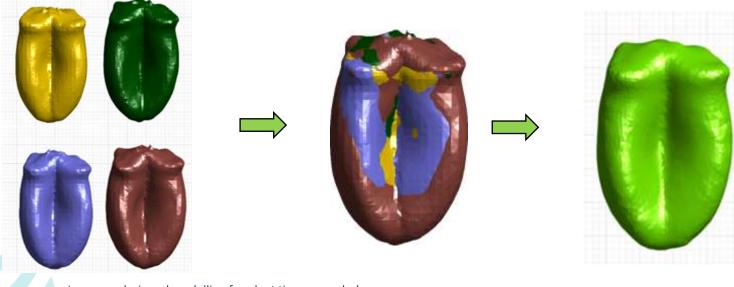




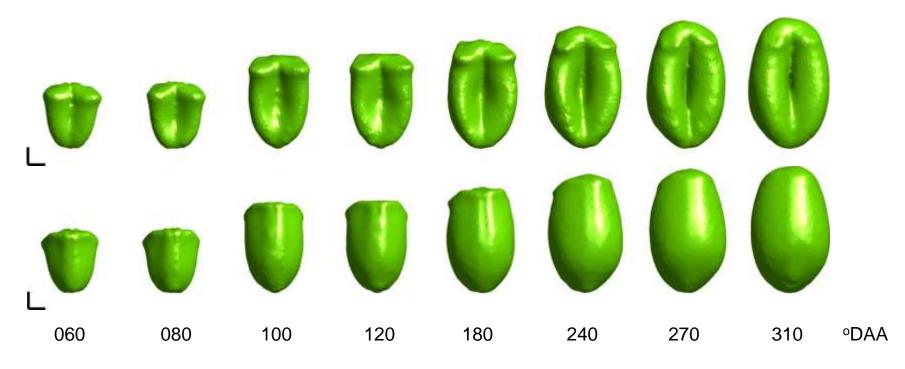
p. 11

Representative shape for each stage

- Computation of average shapes
 - One shape per stage
 - Principle:
 - Global rigid alignment (ICP) + scaling
 - "Least square surface" computation



> Time evolution of average shapes



- Main shape variations are well preserved
- "smoothing" effect

0

INRAØ

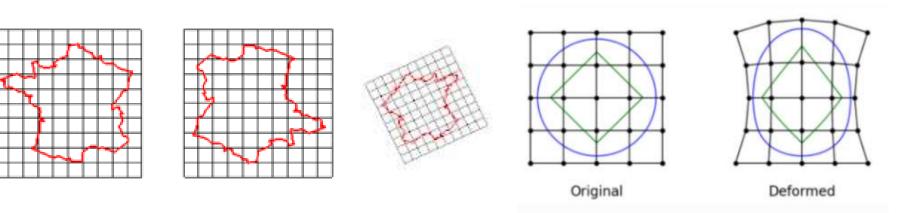
Choice of transform model

- "Rigid" transforms
 - Translation
 - Rotation
 - (uniform) Scaling
 - => Similarity

• Elastic transforms

- Polynomial
- Displacement fields
- Free-form deformation (Bsplines)

• .



INRAØ

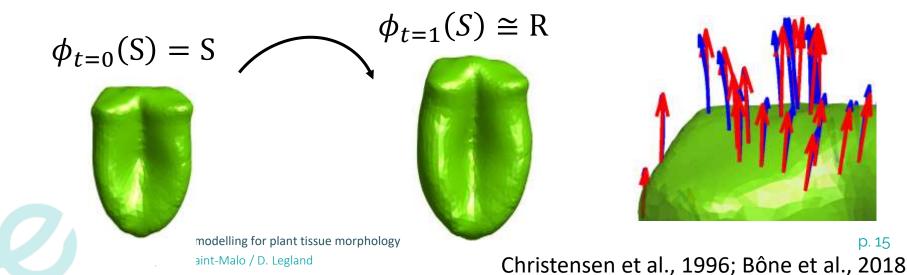
Image analysis and modelling for plant tissue morphology 2024-06-13 / AFH Saint-Malo / D. Legland Rueckert et al., 1999 p. 14 Sánchez Sorzano et al., 2005 (BUnwarpJ)



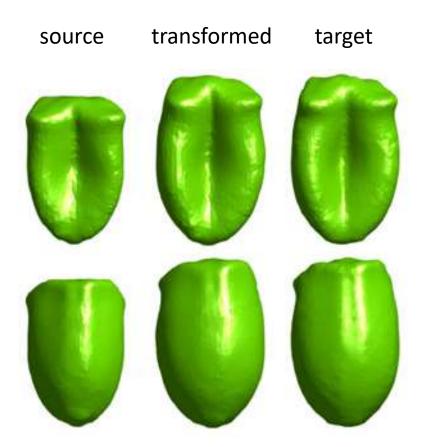
- "Large deformation diffeomorphism metric mapping" (LDDMM) framework
 - Diffeomorphism: smooth & invertible transform
 - Parameterization with "time"
 - t=0: stage *i*
 - t=1: stage *i*+1

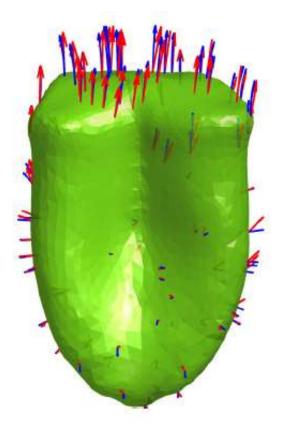


p. 15



Computation of deformations - results



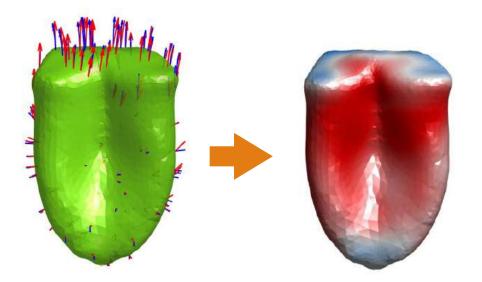


How to analyze a 3D deformation field?





- Aim: relate local deformation to growth
- Several features
 - Local displacement
 - Local derivatives
 - .
- Representation by means of parametric maps





INRAØ

Image analysis and modelling for plant tissue morphology 2024-06-13 / AFH Saint-Malo / D. Legland p. 17 Rueckert et al., 2008



- Computed as the determinant of the Jacobian matrix
 - Logarithmic scale
- Depicts local variation of volume
- Localization mostly in the upper part of the grain

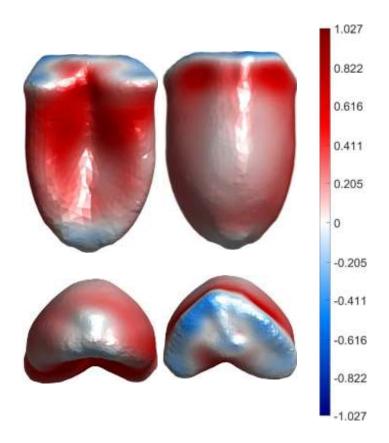
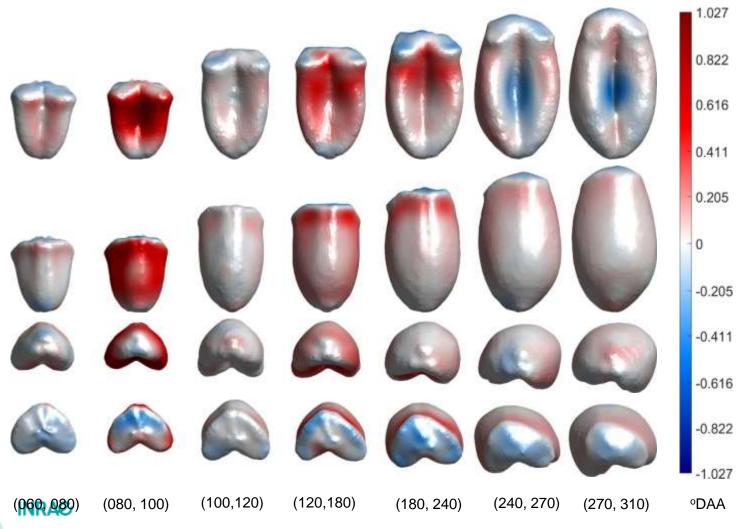




Image analysis and modelling for plant tissue morphologyNote: can be used to check deformation2024-06-13 / AFH Saint-Malo / D. Leglandobtained from expansion microscopy

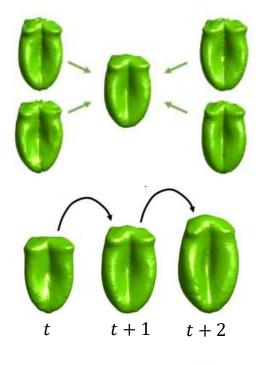


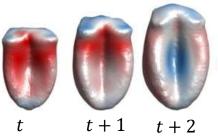


- Computation of a representative "average grain" for each stage
- Computation of the geometric deformations between average grains
- Analysis of deformations
 - Local deformation maps
 - Relate to growth



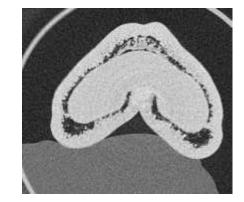
INRA@







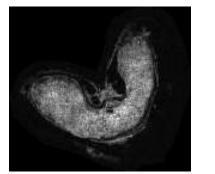
- Fusion of images from different modalities
 - Anatomy (tomography, microscopy...)
 - Composition (microscopy, MSI...)
 - Water mobility (MRI)
 - Mechanical properties (AFM)



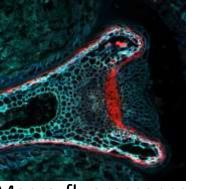
•



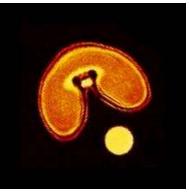
Microscopy



MALDI



Macro-fluorescence

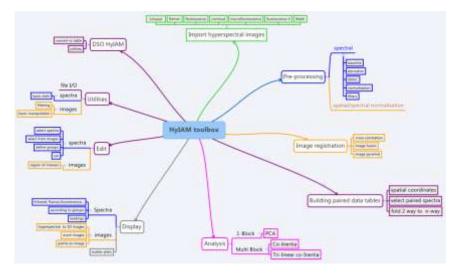


MRI

INRAØ

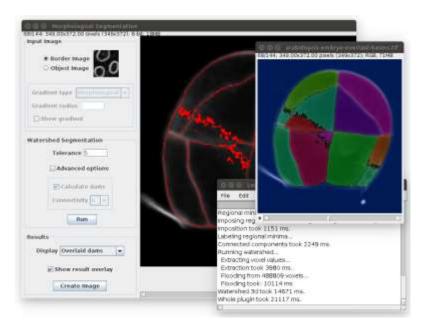
Diffusion / valorization of methodological developments

- Matlab
 - Shape / image registration
 - Granulometry
 - Multivariate image analysis
 - ...



ImageJ / Fiji

 MorphoLibJ: Morphological image processing



INRAØ



Cell Wall team

- A.L. Chateigner
- Thang Le
- M. F. Devaux
- F. Guillon
- M. Lahaye
- C. Alvarado
- S. Durand
- J. Beaugrand
-

GDEC lab

• C. Girousse

BIBS Facility

- H. Rogniaux
- D. Ropartz
- B. Novales
- A. D'Orlando
- ...

SOLEIL Synchrotron

- F. Jamme
- M. Réfrégiers
- A. King
- C. Rivard
- ...

ONIRIS

• M. Hanafi

IJPB

- P. Andrey
- J. Burguet
- E. Biot
- V. Méchin
- M. Reymond

PIAF team

• E. Badel





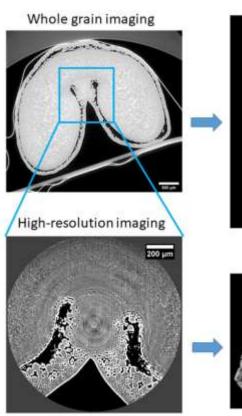


> X-ray imaging at SOLEIL synchrotron

Psiché beamline – Feb. 2018

• Objectives:

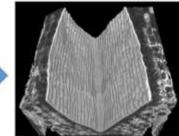
- Imaging of thin tissues
- Imaging at cellular scale
- Explore space x time heterogeneity
- Sampling design
 - Ten developmental stages
 - Two imaging scales
 - Full-grain imaging
 - High resolution imaging



3D rendering



3D rendering

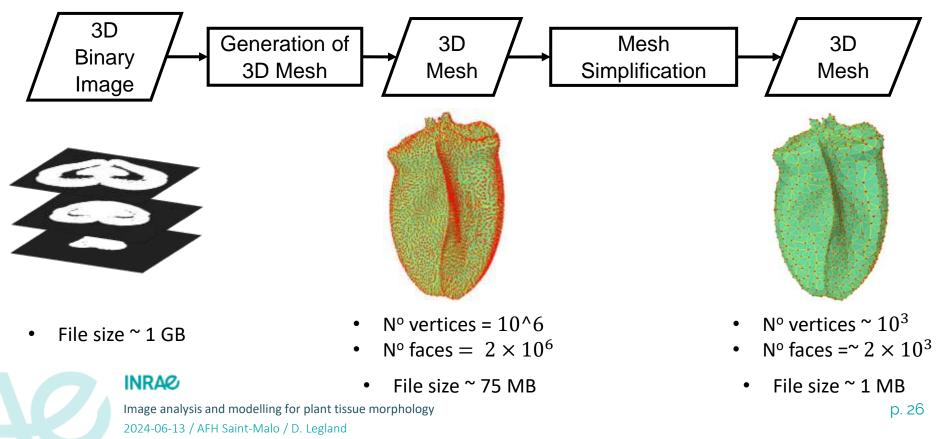


INRAØ

> Conversion into polygonal meshes

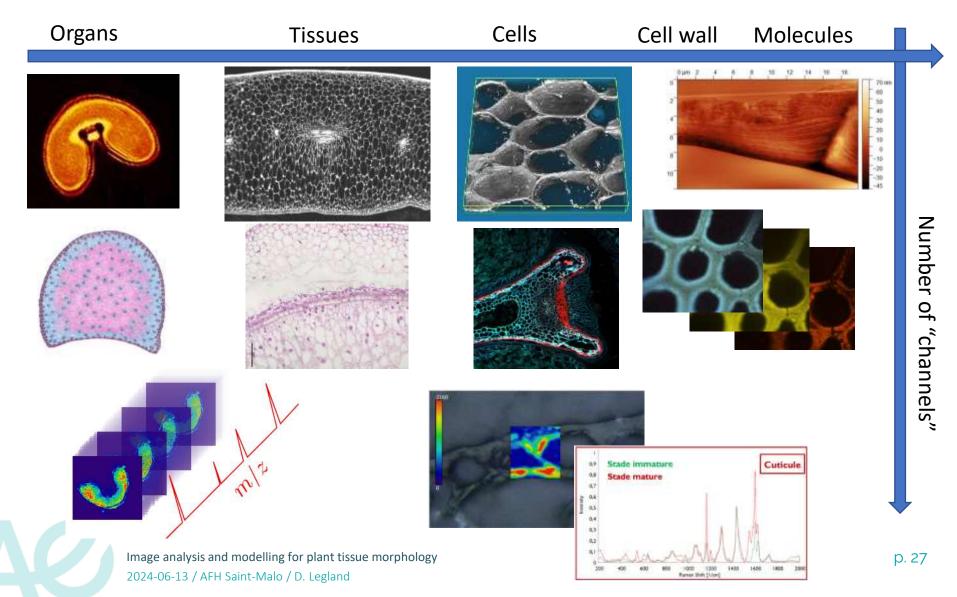
Reduction of computational complexity

Simplification of results interpretation

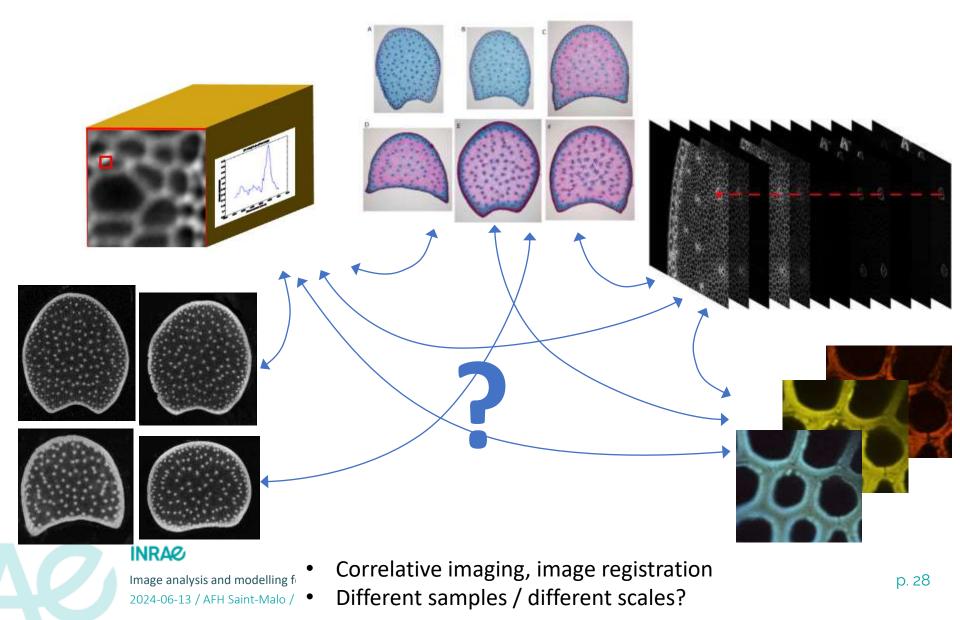


> Multi-scale imaging of plant tissues

Joint investigation of morphology and localized biochemistry

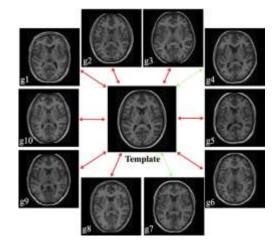


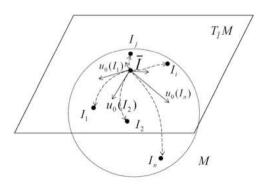
How to relate information obtained from different images / modalities



> A "computational anatomy" approach

- Originally developed in a context of medical imaging
- Computation of a reference shape from a collection of individual shapes
 - Description of shape population ("shape space")
 - Comparison of different populations
 - Integration of localized quantitative data obtained on objects with different shapes



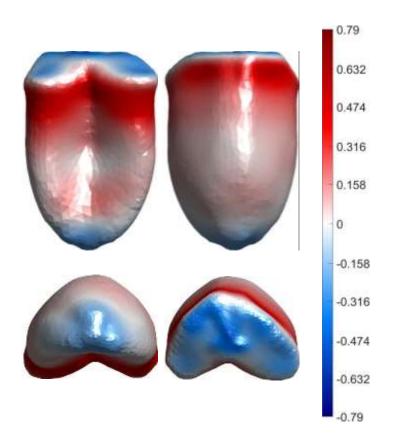


p. 29 Grenander & Miller, 1998





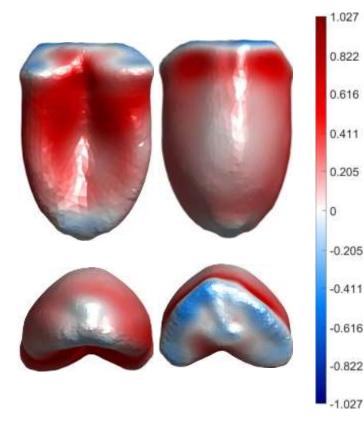
- Computed as the (3,3) coefficient of the Jacobian matrix
 - Logarithmic scale
- Depicts relative elongation in vertical direction
- Results seems similar to global scaling



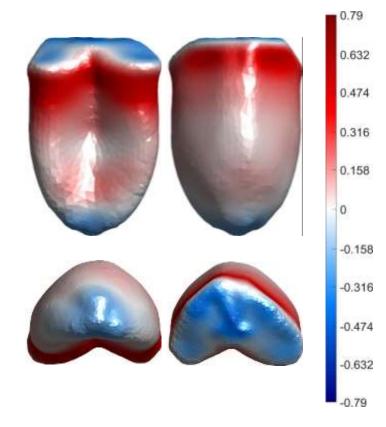


Slobal vs vertical growth

• Local scaling map $\Delta LD(x)$



Vertical growth map $\Delta VG(x)$



 $\Delta LD(x) > 0$ = Increase in volume

 $\Delta VG(x) > 0$ = Vertical expansion in shape

- $\Delta LD(x) > 0$, $\Delta VG(x) > 0$: Mostly elongation
- **INRAO** $\Delta LD(x) > 0$, $\Delta VG(x) \simeq 0$: Mostly thickening