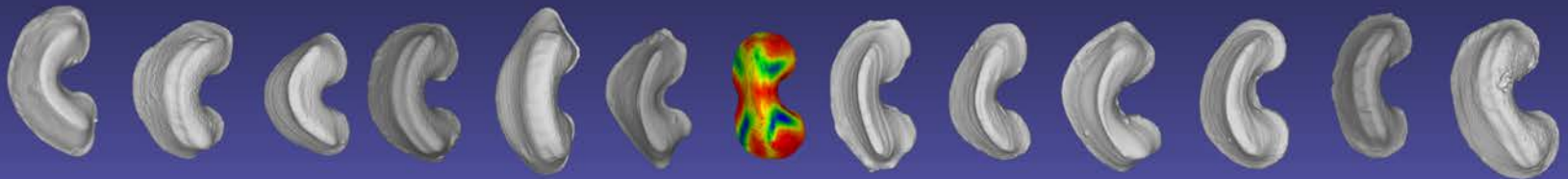


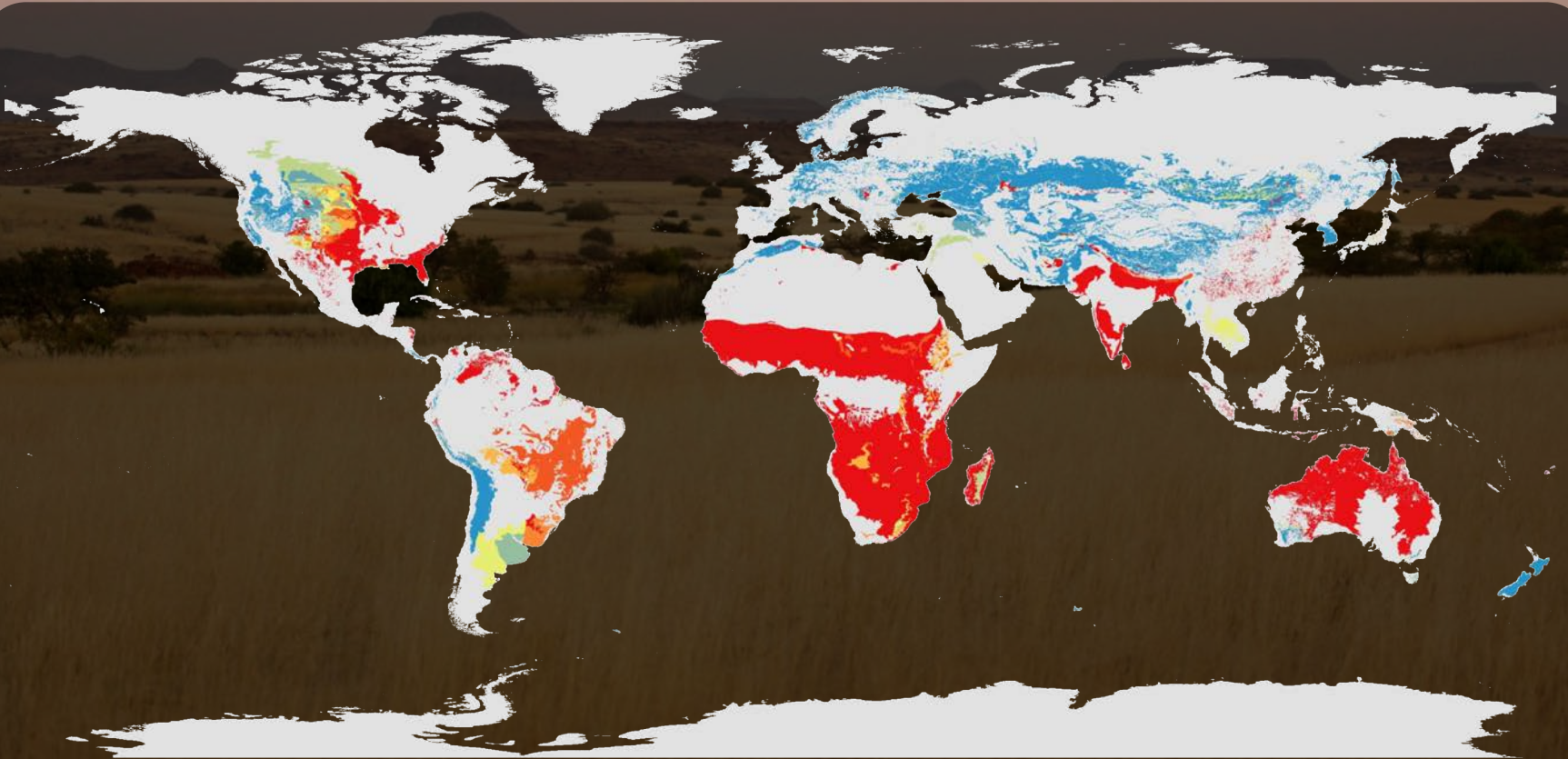
# Shedding light on the evolution of grasses and grasslands through automated, quantitative imaging analyses of plant silica microfossils

Caroline A.E. Strömberg  
Department of Biology & Burke Museum  
University of Washington  
Seattle, WA



# Grasslands are ecologically vital

- Grassy biomes make up >40% of Earth's land surface



Color = grass-dominated habitats

Lehmann et al. (in prep.)

# When and how did grassland ecosystems come to be?

- When did the grass family first originate and diversify?
- When did grasses become ecologically dominant, forming grasslands?

# Direct evidence for past grasslands

- Grass mesofossils and pollen are rare until the late Miocene—and often hard to interpret taxonomically



Pollen



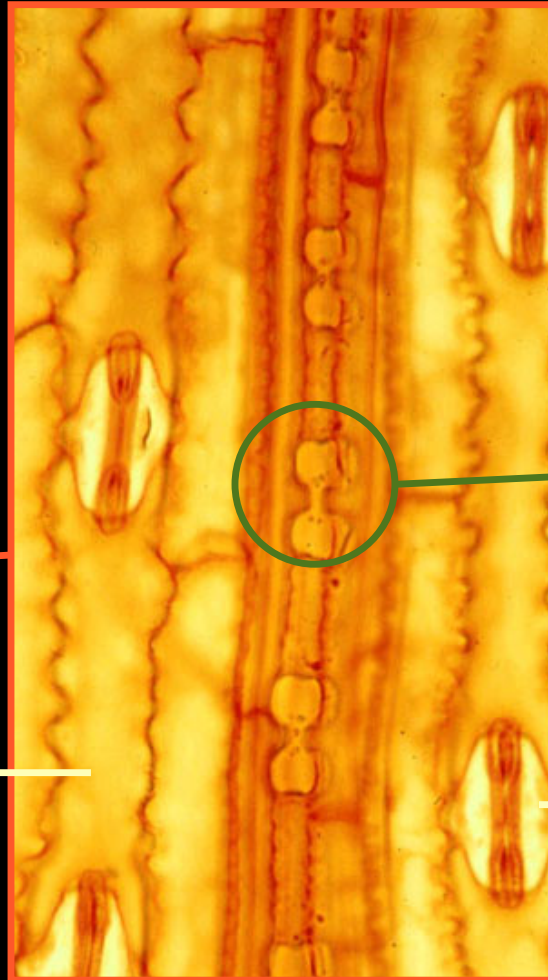
Leaves



Fruit

# Phytoliths (*plant silica*)

Grass epidermis:



epidermal  
cell

stomata

Grass phytolith  
in sediment:



10  $\mu\text{m}$



# Phytoliths (*plant silica*)

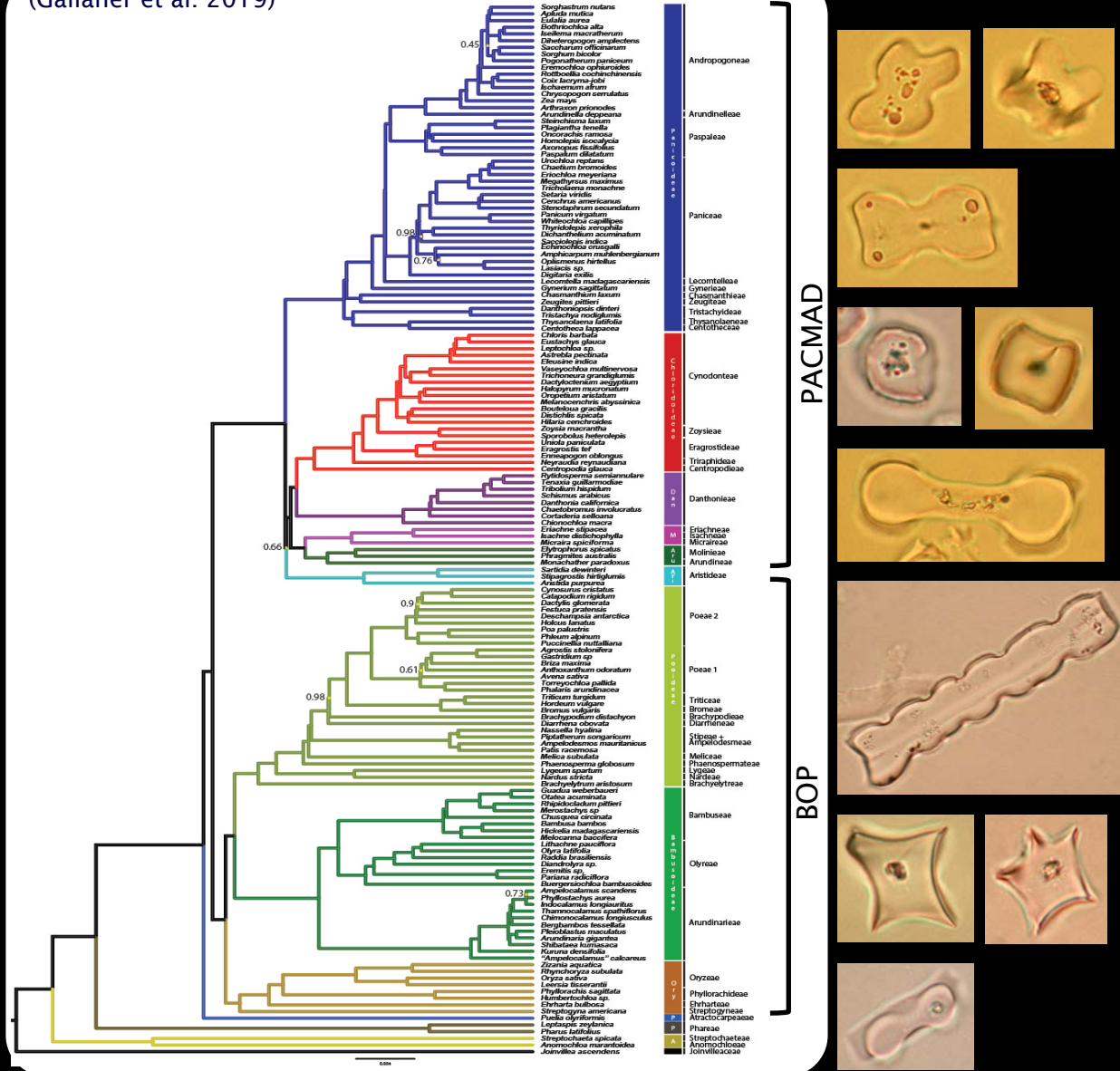
- Taxonomically useful variation within Poaceae (grass family)

**PACMAD:**  
 Panicoideae  
 Arundinoideae  
 Chloridoideae  
 Micrairoideae  
 Aristidoideae  
 Danthonioideae

**BOP:**  
 Bambusoideae  
 Oryzoideae  
 Pooideae

Early-diverging grasses

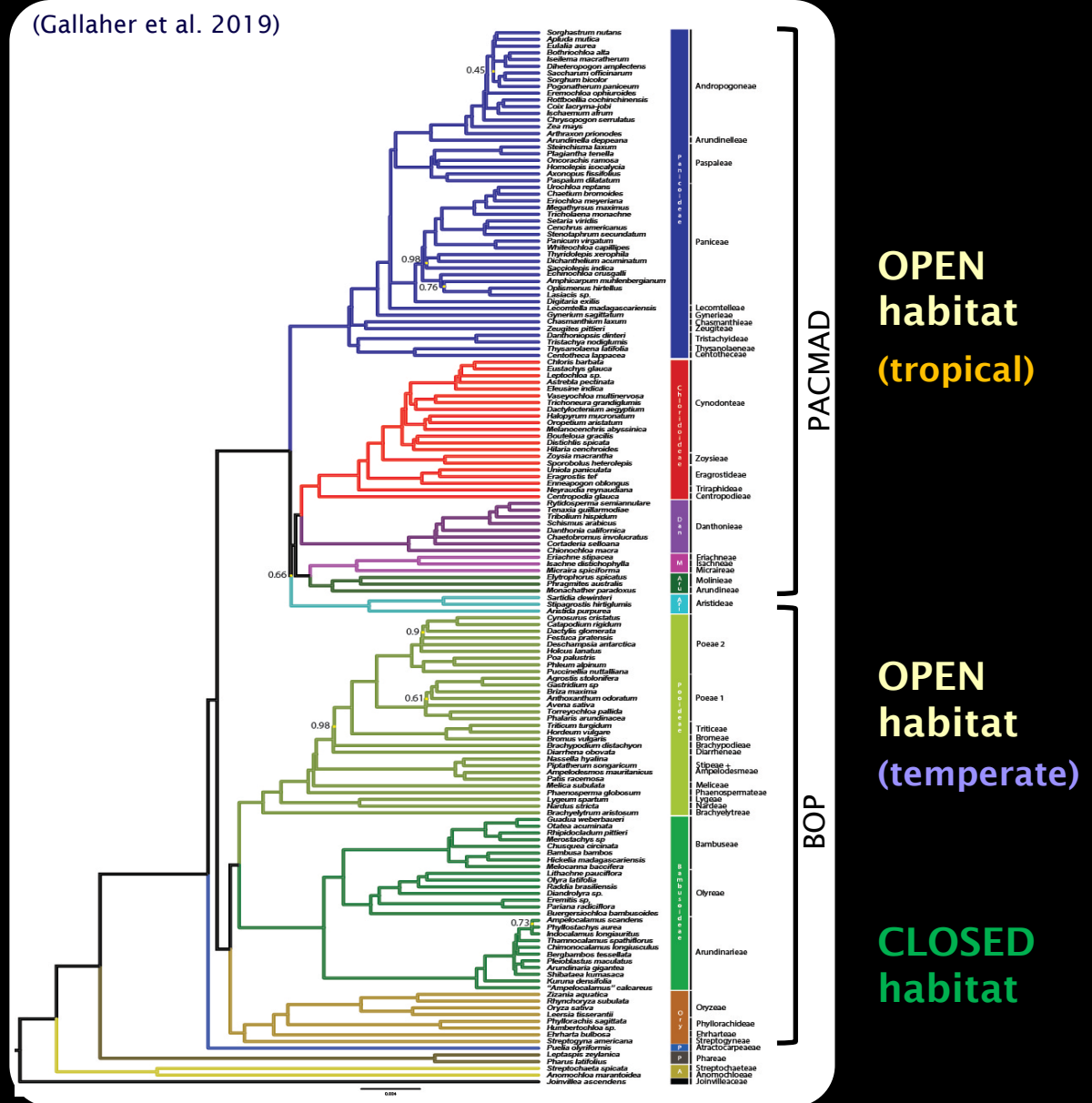
(Gallaher et al. 2019)



# Phytoliths (*plant silica*)

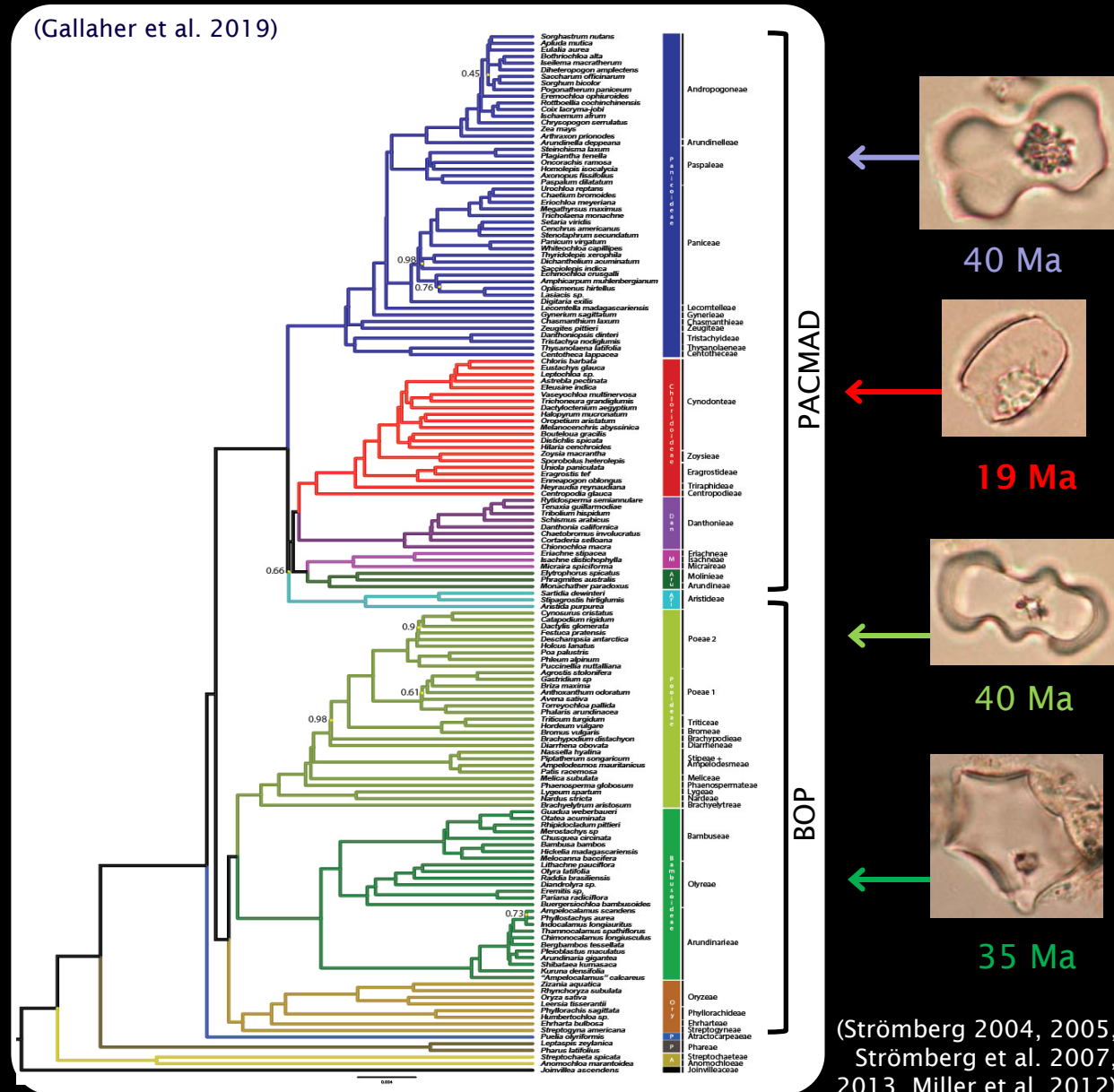
- Taxonomically useful variation within Poaceae (grass family):
  - Diversification of ancient grass lineages
  - Ecology of past grass communities

(Gallaher et al. 2019)



# Radiation of open-habitat grasses

- Fossil phytolith morphotypes (Americas, Eurasia):  
 → Open-habitat grasses diversified by 40 Ma





# Grassland evolution in North America

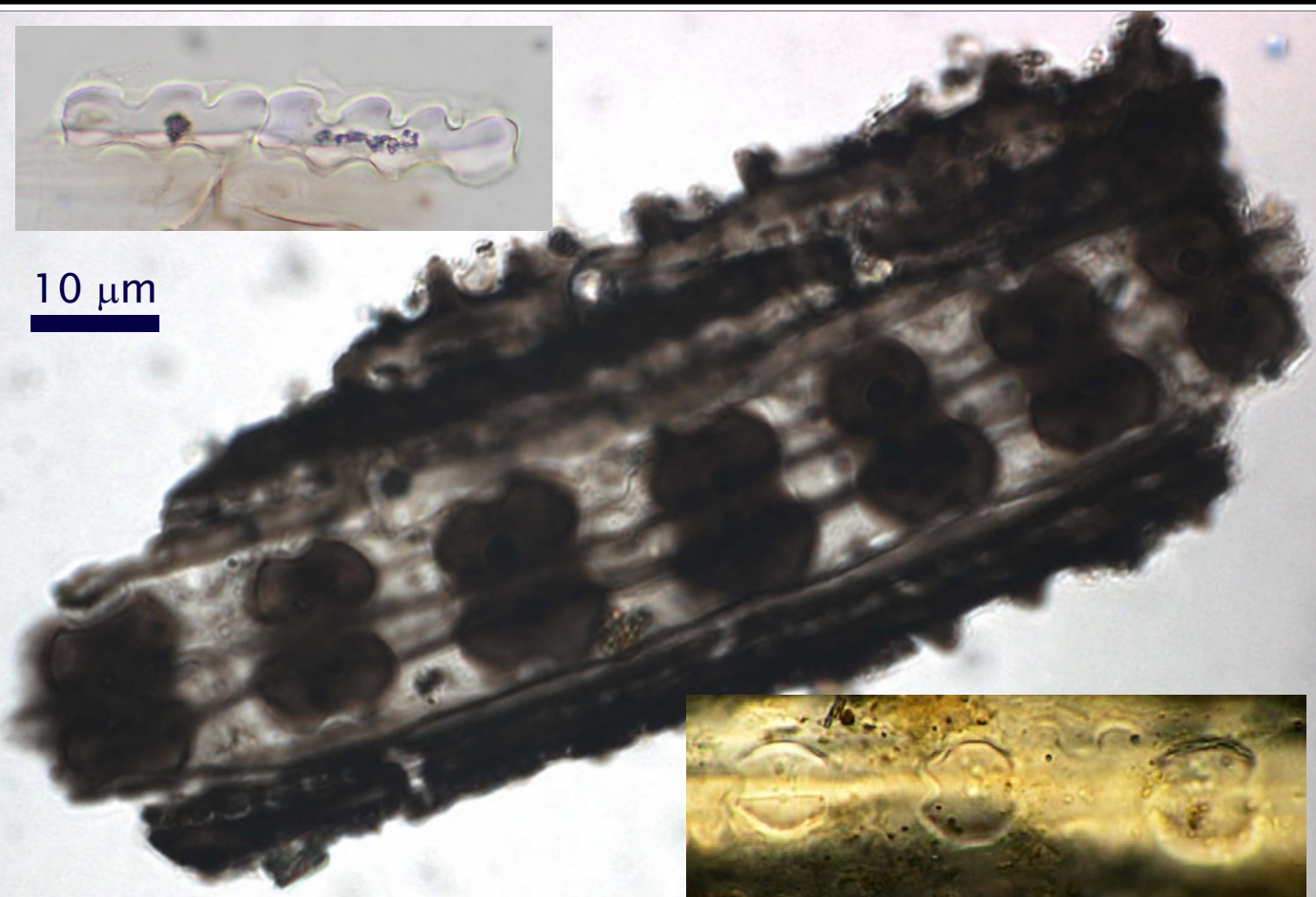
- Earliest (early Miocene) grasslands were dominated by cool-temperate stipoid pooids
- Tropical, dry-adapted ( $C_4$ ) chloridoids spread during the latest Miocene



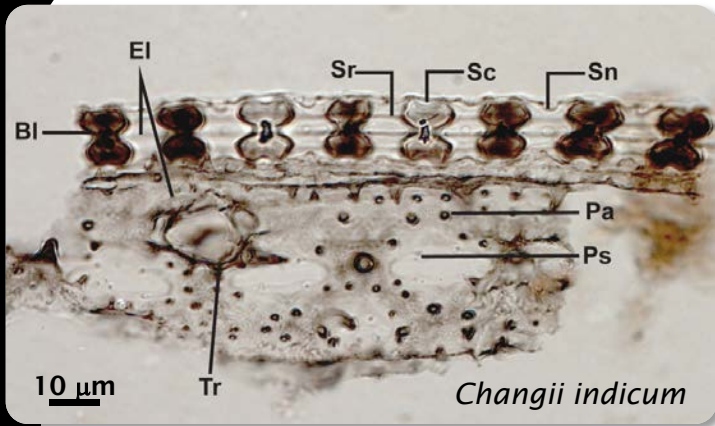
# Early grass diversification

- Phytoliths (+cuticle) from Late Cretaceous dinosaur coprolites and sediment, central India

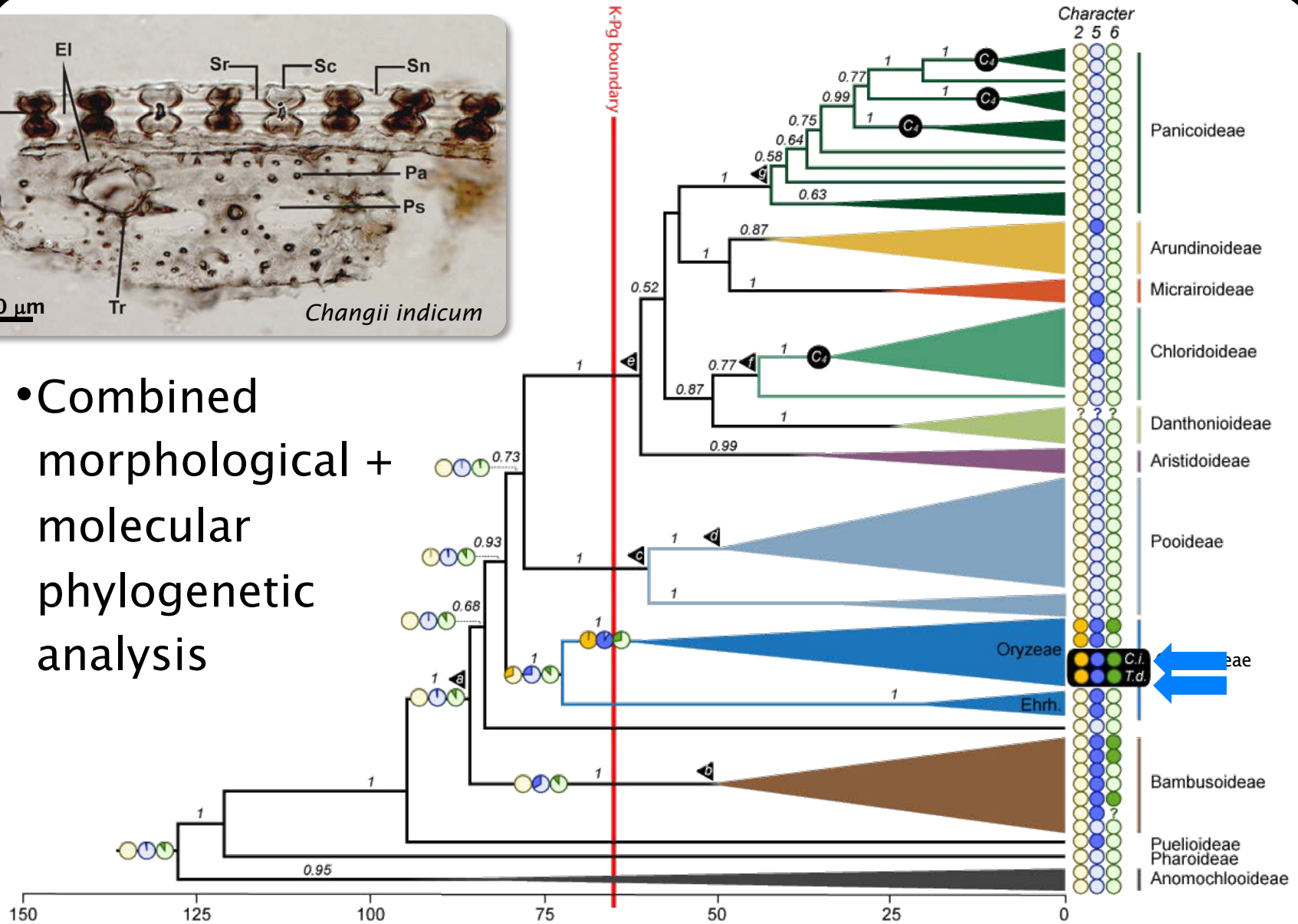
Prasad et al. (2005, 2011)



# Early grass diversification

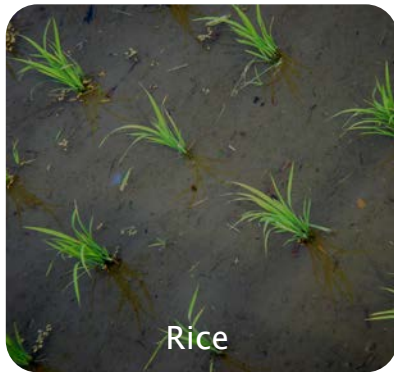


- Combined morphological + molecular phylogenetic analysis

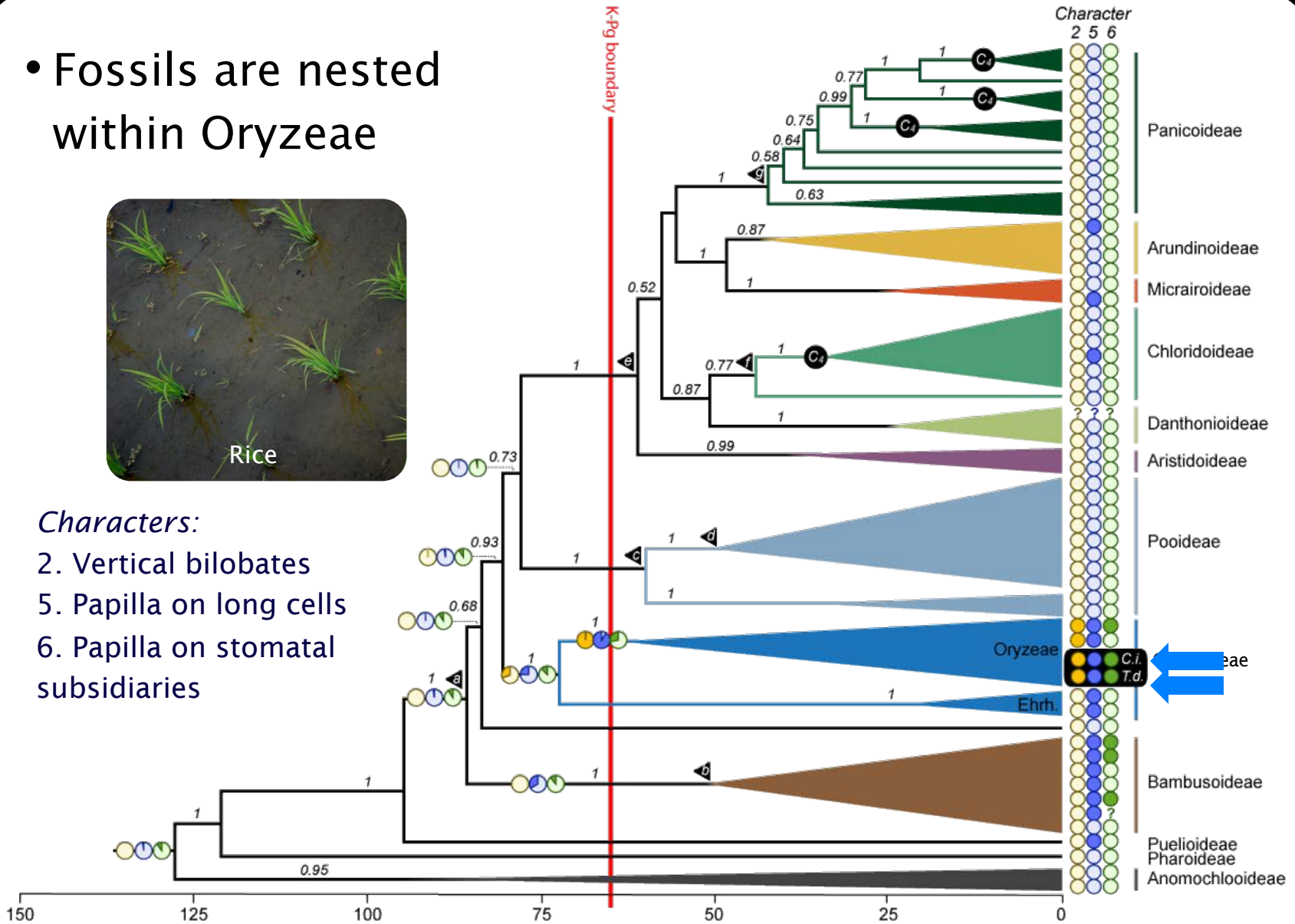


# Early grass diversification

- Fossils are nested within Oryzeae

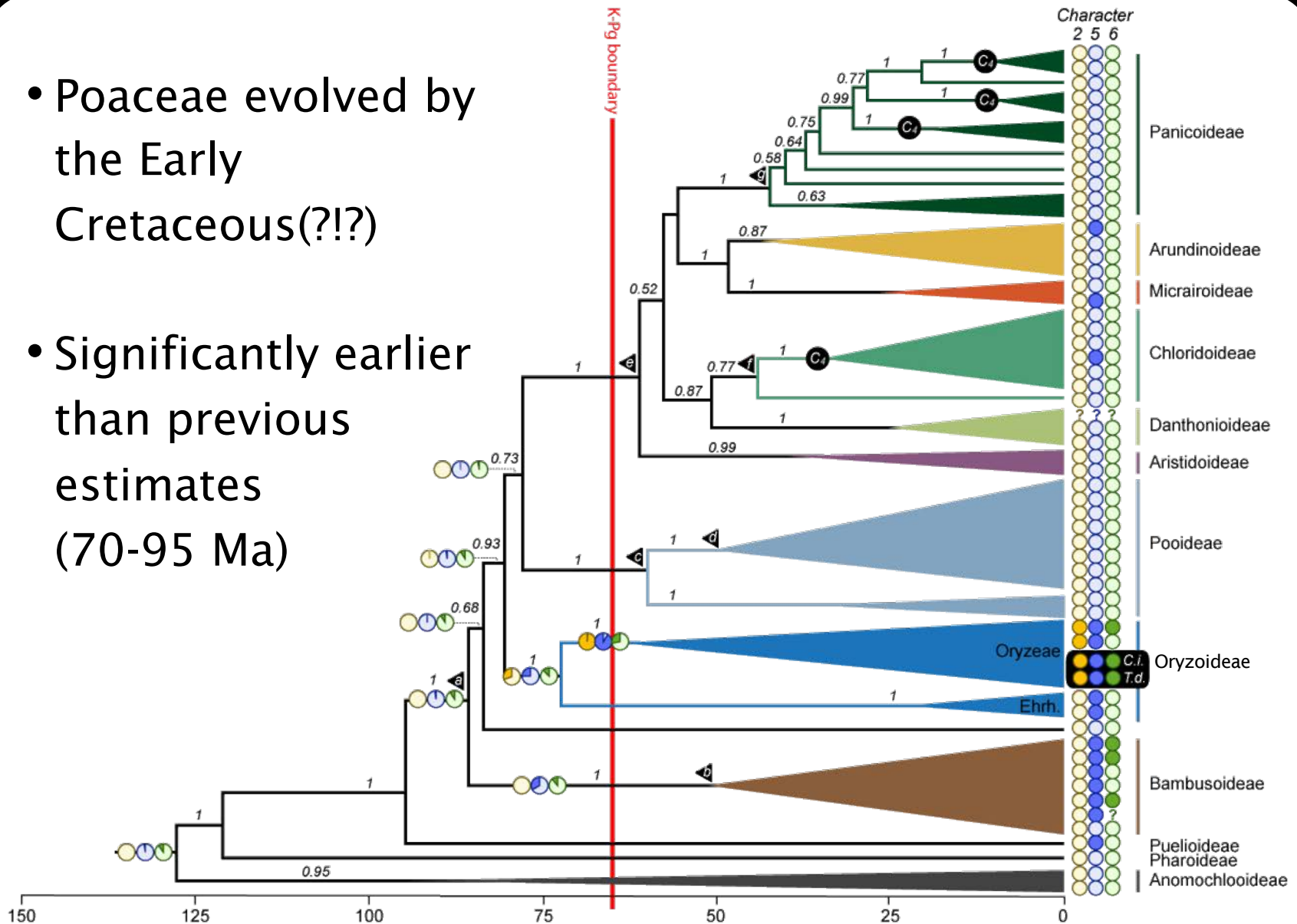


- Characters:
- 2. Vertical bilobates
  - 5. Papilla on long cells
  - 6. Papilla on stomatal subsidiaries



# Early grass diversification

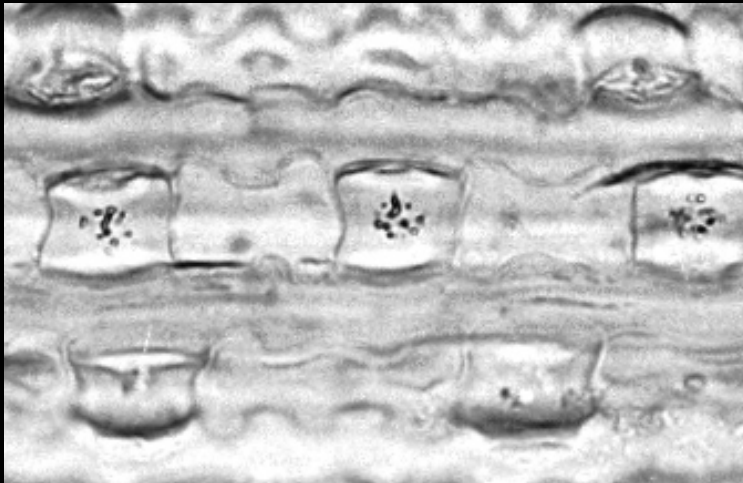
- Poaceae evolved by the Early Cretaceous(?!?)
- Significantly earlier than previous estimates (70-95 Ma)



# Problem with placing fossil phytoliths in Poaceae phylogeny

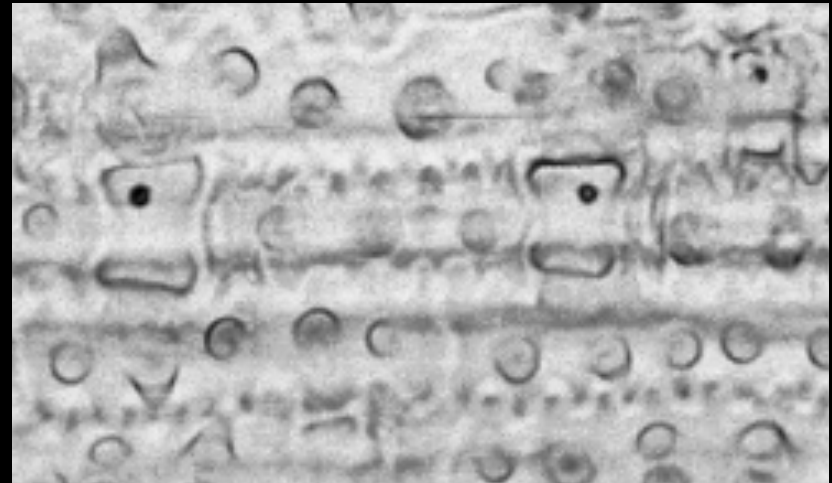
- **Problem 1:** Description and classification of grass phytoliths qualitative/semi-quantitative, 2-D, and subjective

“saddle-shaped”



*Chloris*

Dry-adapted, C<sub>4</sub> open-habitat grass



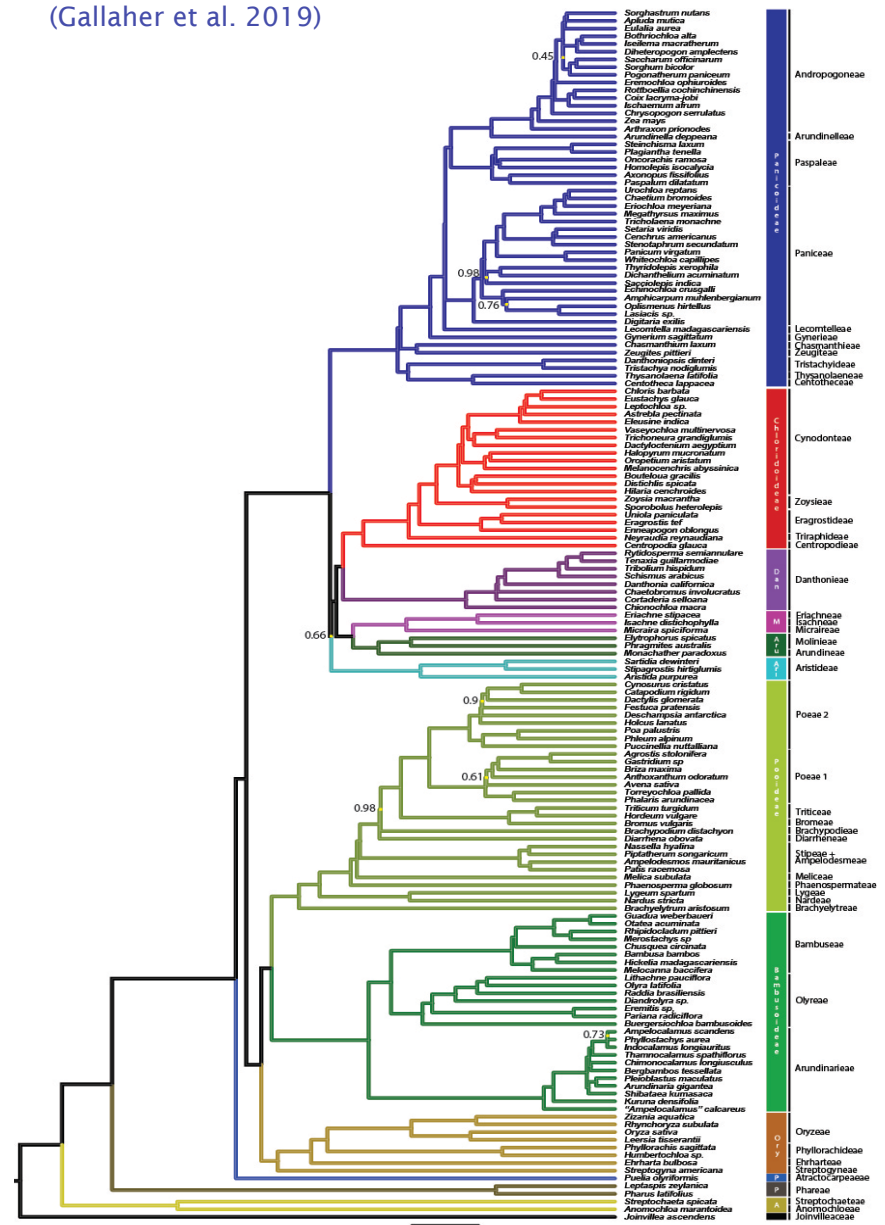
*Sinobambusa*

Mesophytic, tree-forming C<sub>3</sub> bamboo

# Problem with placing fossil phytolites in Poaceae phylogeny

- Problem 2: Based on outdated grass taxonomy
- Recent phylogenies have dramatically changed our understanding of Poaceae relationships

(Gallaher et al. 2019)



*Needed:*

**Precise, objective (and phylogenetically based) way to place grass phytoliths taxonomically—a “phylogenetic key”**

Grass silica short cell phytoliths (GSSCP)





# Creating a phylogenetic key to grass phytolith shape

- Approaches to GSSCP identification/placement
- Properties of GSSCP
- Building a 3-D shape key to GSSCP
- Preliminary application to the Late K-Paleocene of India
- GSSCP and Machine Learning & Computer Vision

# Creating a phylogenetic key to grass phytolith shape

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# Team GRASS



UW Undergraduates

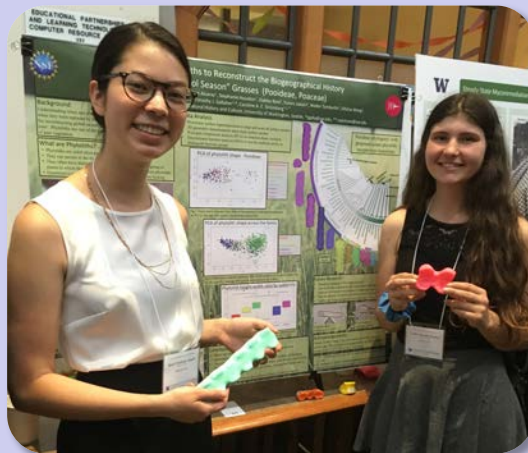


Postdoc Tim Gallaher

Lynn Clark,  
Iowa State  
University



Ashly Senske Walter Tomberlin



Kari Jessett

Callie Zender



Sultan Akbar



Grad student  
Jessica Perry



Kevin Jamieson, UW

Anna Schorr, Nik Pershing,  
Elie Aboulafia, Brittany  
McManus, Casey O'Keefe,  
Claire Marvet, Brian Connor,  
Maura Kilpatrick, Yutaro  
Sakairi, Angel Alcaraz,  
Oakley Reid, Gwen Xiao,  
Claire Grant, Alex  
Arrendale, Sophia Druet,  
Tim Novak, Jessica O'Hanlon



# Approaches to GSSCP identification/ placement

1. Landmark based 3-D surface geometric morphometrics
  - + 3-D shape
  - landmarks, time consuming
2. Machine Learning and Computer Vision
  - + regular brightfield images, no landmarks needed
  - 2-D shape, “black box”



# Creating a (phylogenetic) key to GSSCP 3-D shape

## *Goals:*

- Measure 3-D shape of GSSCPs using geometric morphometrics
- Map 3-D shape onto current phylogeny
- Correlate with ecological and physiological characters



*→ Trace evolution of GSSCP shape and size across Poaceae*

*→ Establish GSSCP shape/size diagnostic of particular clades/ecologies/physiologies*



# Creating a phylogenetic key to grass phytolith shape

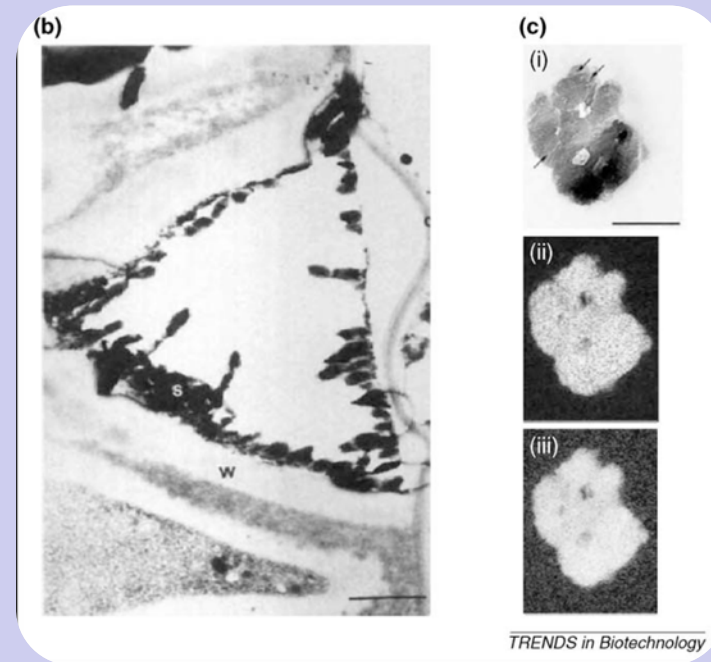
- Approaches to GSSCP identification/placement
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# Unique properties of phytoliths

1. Phytoliths consists mainly of  $\text{SiO}_2$  (66–91%), organic carbon OC (1–6%),  $\text{H}_2\text{O}$  (0–11%), Al (0.01–4.55%), and Fe (0–2.1%)

→ *Phytoliths do not auto-fluoresce and do not readily stain—or stain evenly*

Developing phytolith

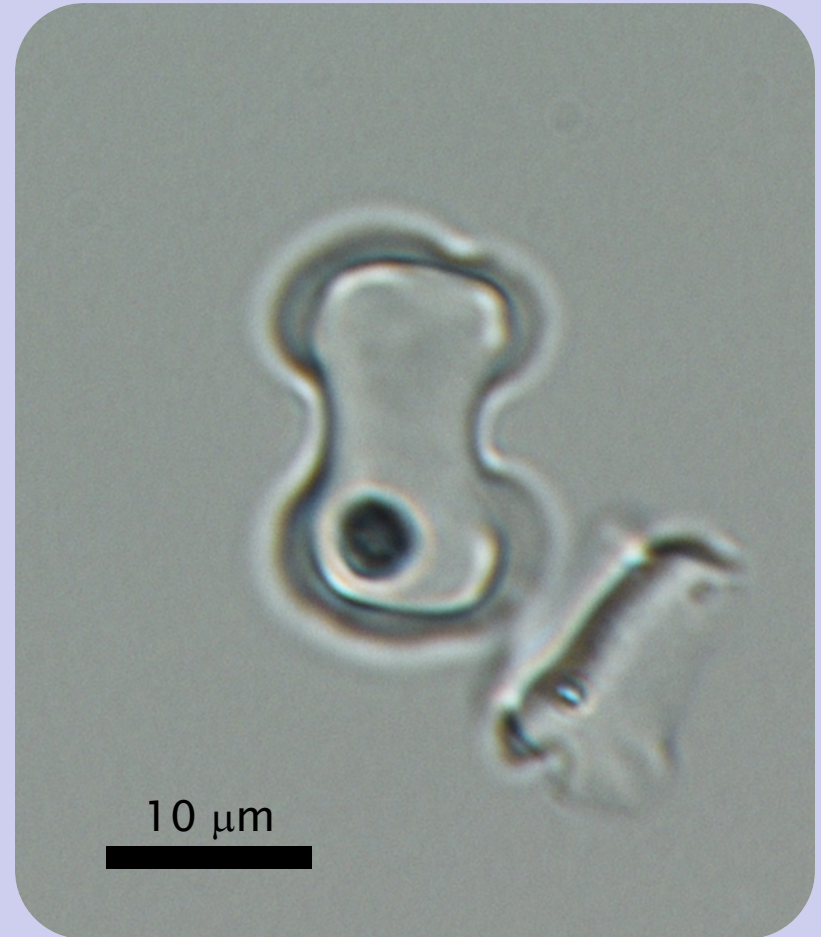


(Neethirajan et al. 2009)

# Unique properties of phytoliths

2. GSSCP phytoliths are small  
(~7-40 micrometers)

→ *Resolution of e.g., micro-CT (100-200 micrometers) is not fine enough*

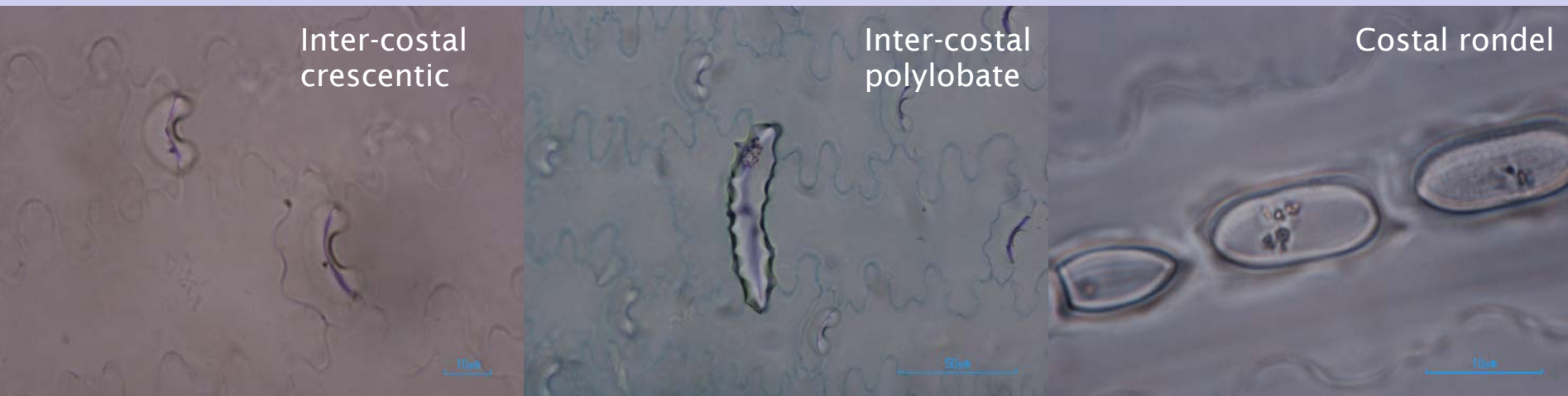




# Unique properties of phytoliths

3. Most grass species make  $\gg 1$  type of GSSCPs

*Anomochloa marantoidea*

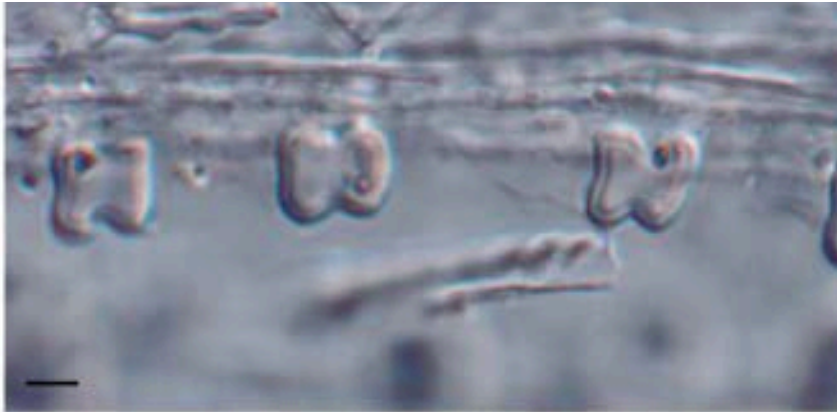


→ *GSSCPs have to be studied like assemblages*

# Unique properties of phytoliths

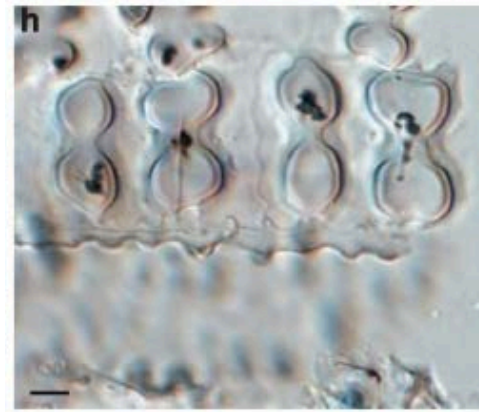
4. Similar GSSCP shapes can be oriented differently in the tissue in different species

*Potamophila*



Axial bilobate

*Zizania*

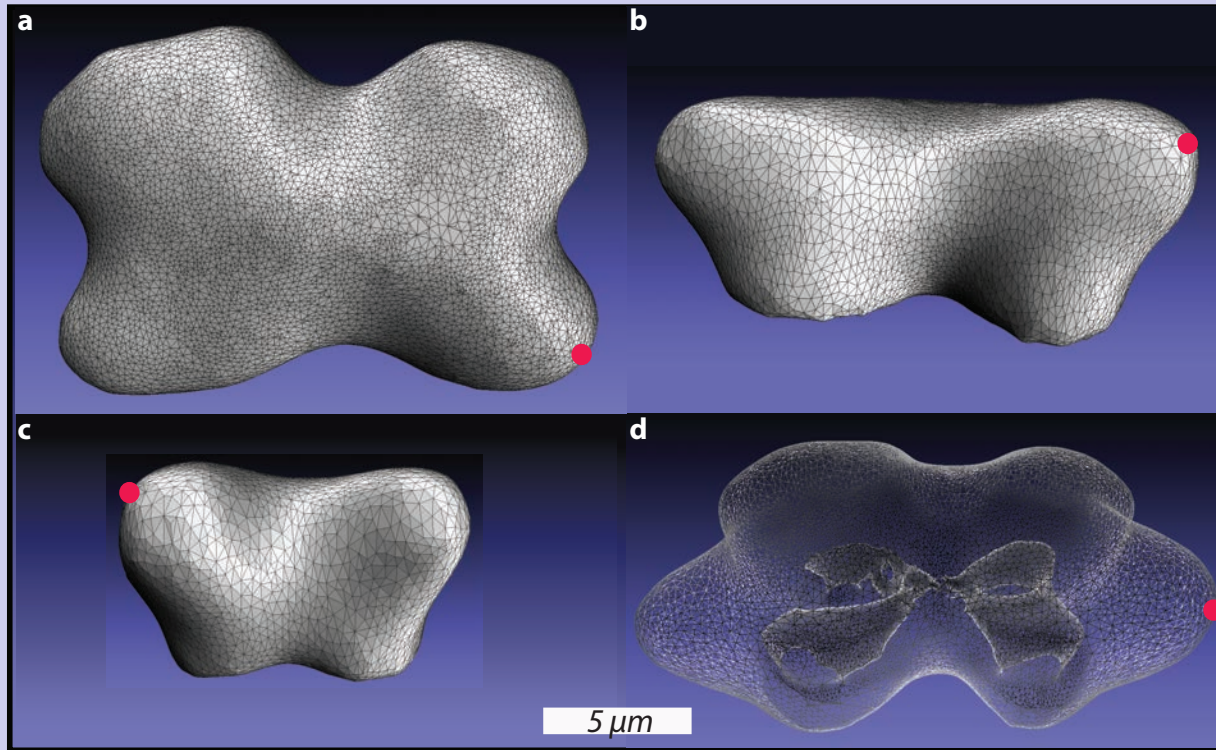


Transverse bilobate

→ *GSSCPs have to be studied in situ*

# Unique properties of phytoliths

5. GSSCP have few good landmarks (homologous points)



→ *Aligning GSSCPs relies on a few (type 3) landmarks only*

# Creating a phylogenetic key to grass phytolith shape

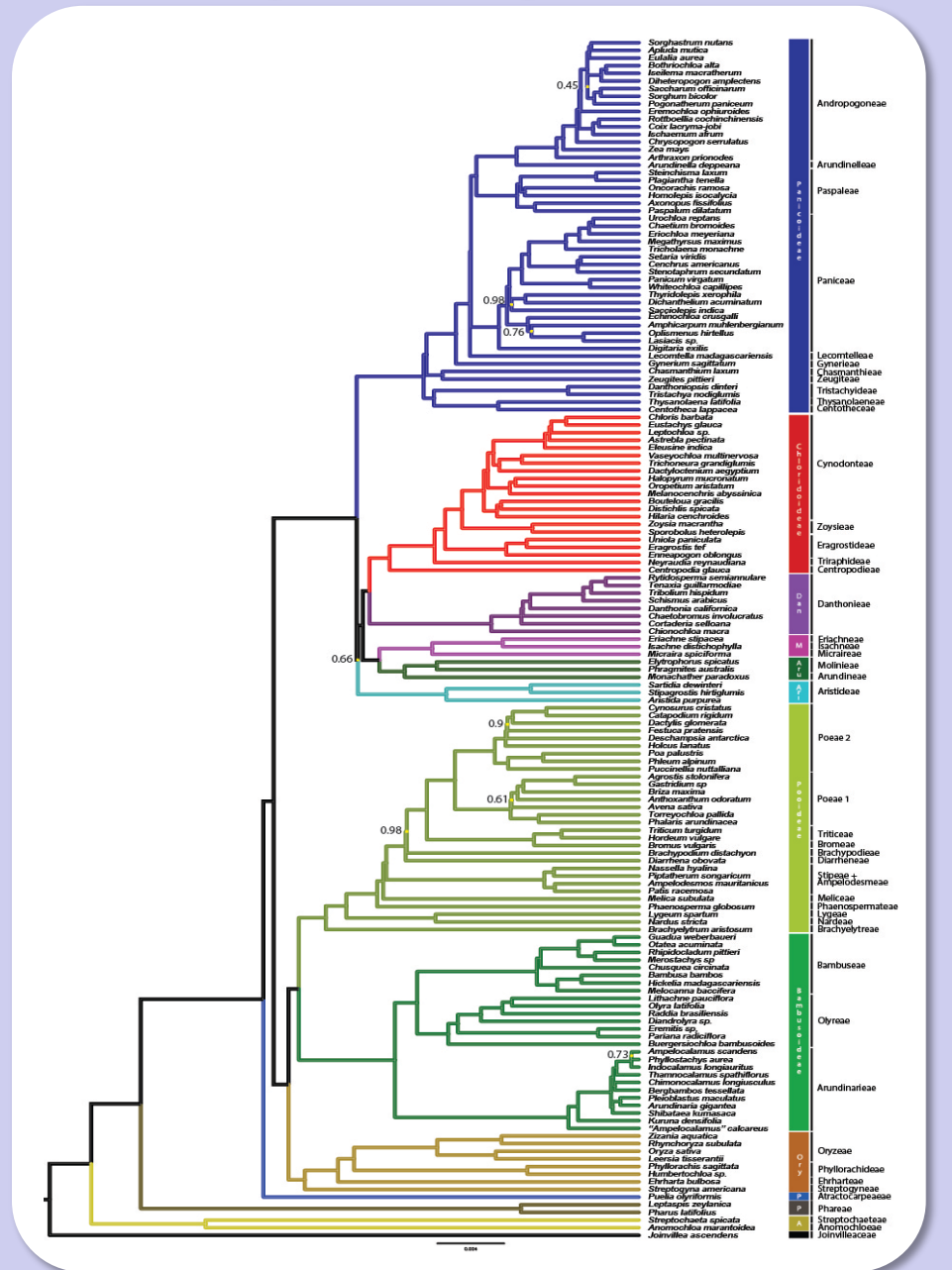
- Approaches to GSSCP identification/placement
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# Building a 3-D GSSCP shape key

## Taxa sampled:

- >200 grass genera from all Poaceae subclades
- Leaf material

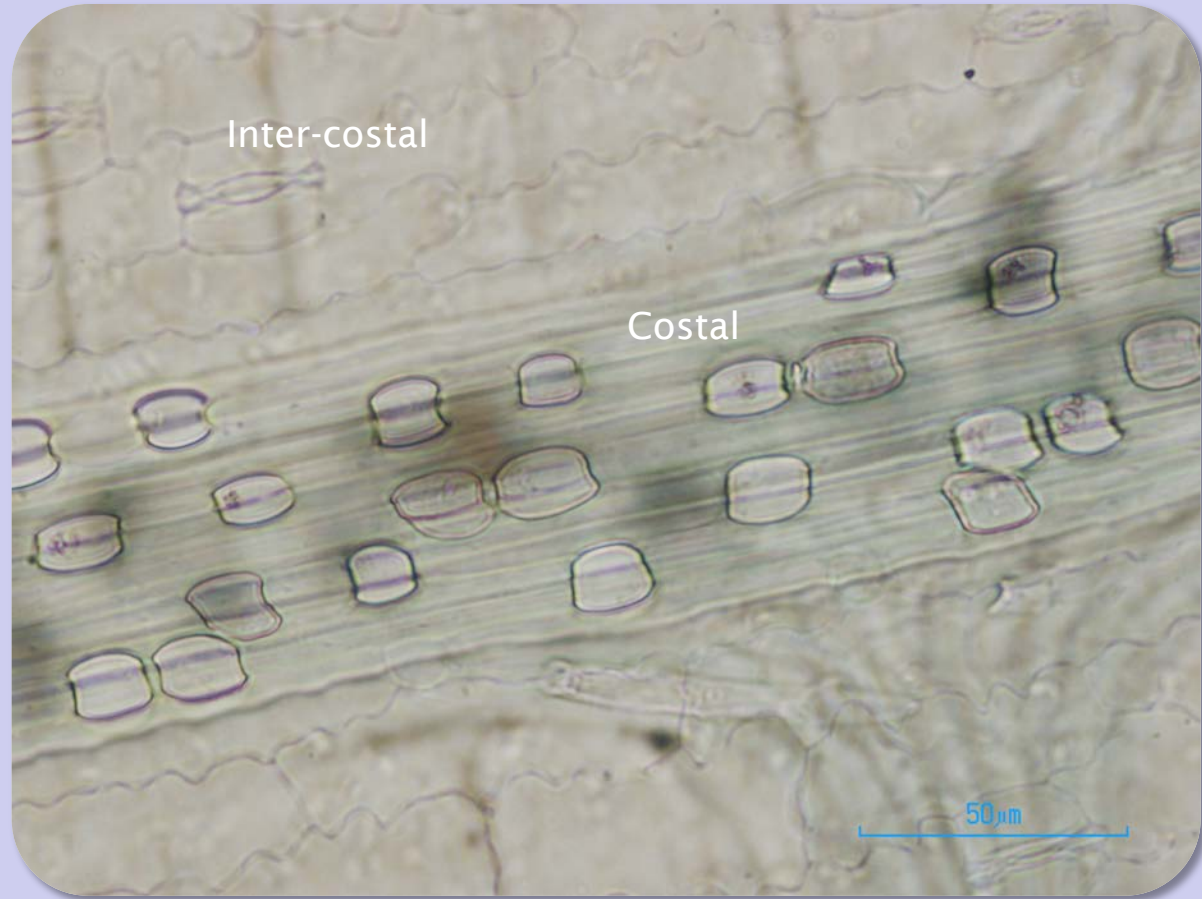
(Gallaher et al. 2019)



# Building a 3-D GSSCP shape key

## *Data collected:*

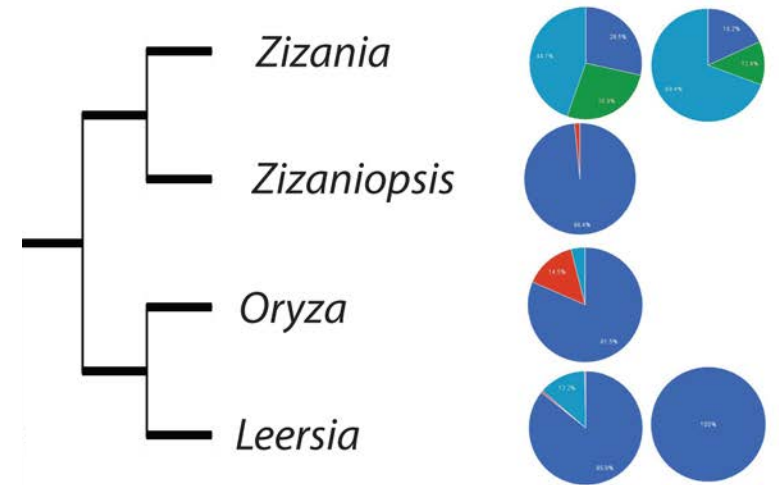
- Orientation and distribution of GSSCP shape



# Building a 3-D GSSCP shape key

## *Data collected:*

- Orientation and distribution of GSSCP shape
- Relative abundances of main morphotypes in GSSCP assemblages extracted from leaves



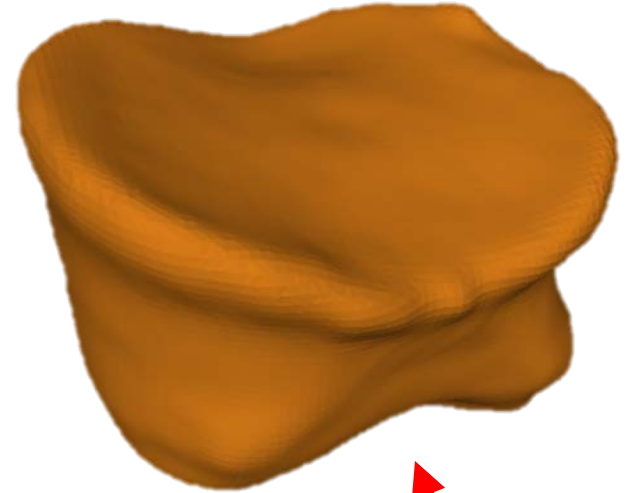
- 200 GSSCP /sample



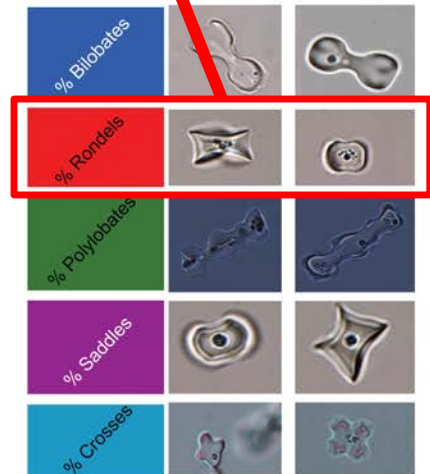
# Building a 3-D GSSCP shape key

## *Data collected:*

- Orientation and distribution of GSSCP shape
- Relative abundances of main morphotypes in GSSCP assemblages extracted from leaves
- 3-D shape within each main morphotype using confocal microscopy of extracted GSSCPs

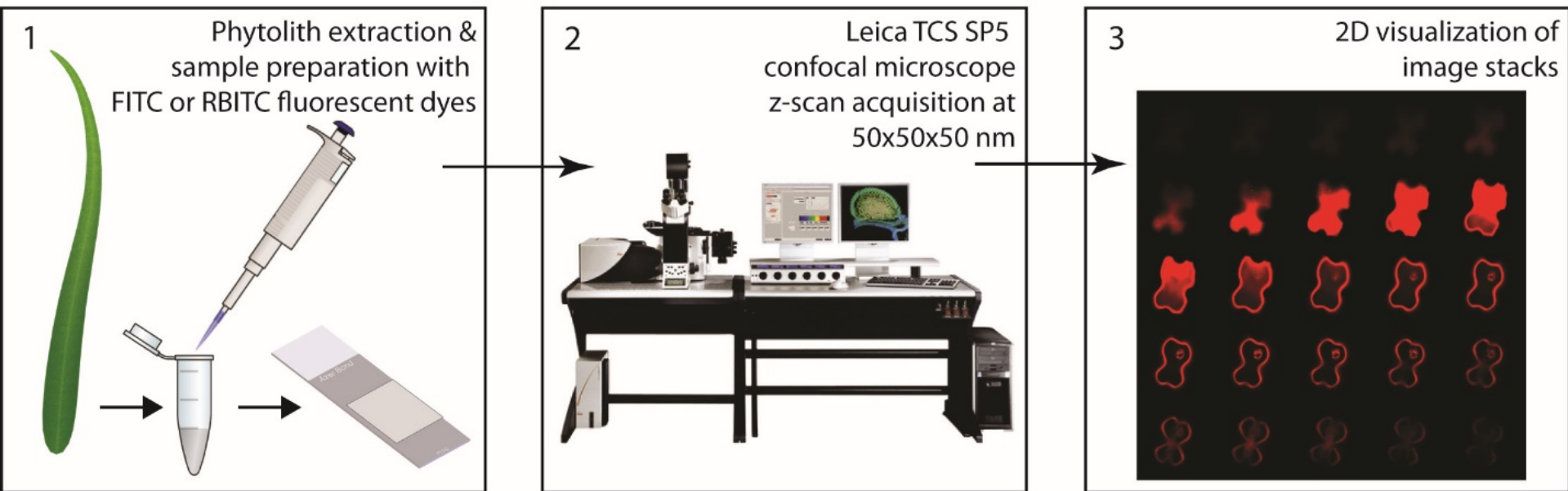


- >10 specimens /GSSCP morphotype





# 3-D data workflow: *Image acquisition*



- Detailed workflow protocol to ensure consistency

Protocols for imaging 3D Phytoliths - Stromberg Lab.  
Updated: Feb. 8, 2017

## Making Slides for Confocal Microscopy

### LABELING

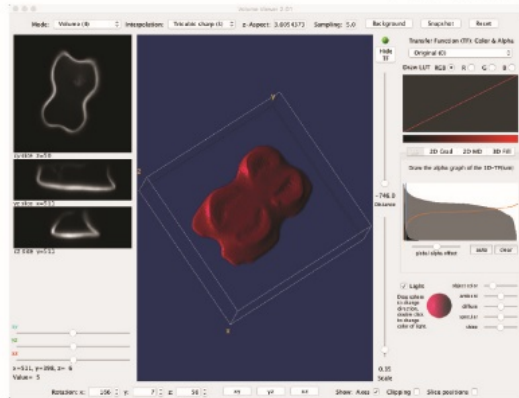
- Label microcentrifuge tubes (1-8 or A-H, etc...) and write sample information in the project notebook

### STAINING/MOUNTING

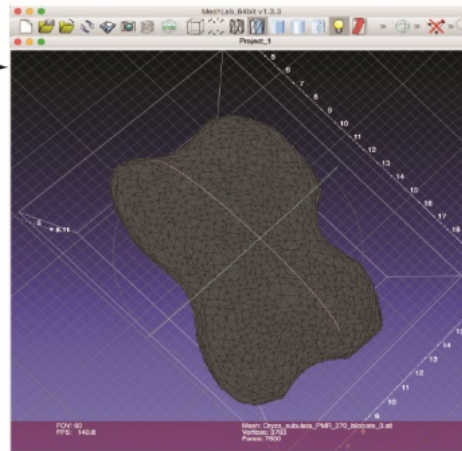
- **Gently vortex vials with extracted phytoliths (in 95% ETOH).**
  - Allow particles to settle (**about 5 min**).
  - With a (p1000) micropipette remove 200 microliters (ul) of solution from the bottom of the glass vial and add it to a 0.7 ml microcentrifuge tube.
- Spin an even number of tubes in the microcentrifuge for 5 min.
- Pipette out and discard supernatant.
- **Add 200 ul of 10% detergent solution (Wash1)**
  - Shake on vortexer (set at 6) with microtube attachment for 10 min.
  - Spin for 5 min. Pipette out and discard supernatant.

# 3-D data workflow: *Image processing and analysis*

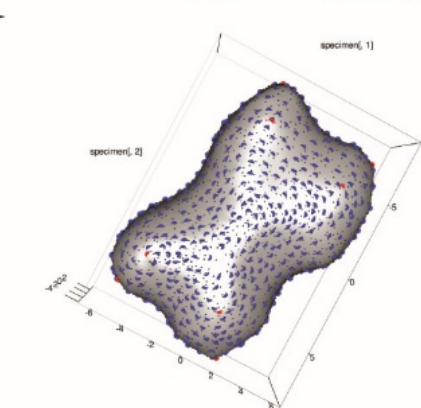
4 Thresholding, background subtraction and image segmentation  
FIJI/IMAGEJ



5 Mesh simplification, cleaning and correction  
MeshLab 1.3.3



6 Digitization of fixed and sliding landmarks and geometric morphometric analysis  
Geomorph 2.0



- Converting image stacks into 3-D models of GSSCPs
- Fit 6 (type 3) landmarks and 1,000 sliding surface semi-landmark points
- Transform and align meshes using Procrustes superimposition to remove size

# 3-D data workflow: *Outcomes*

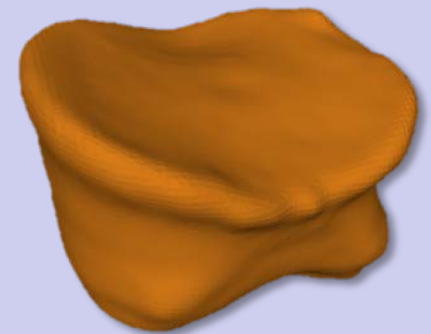
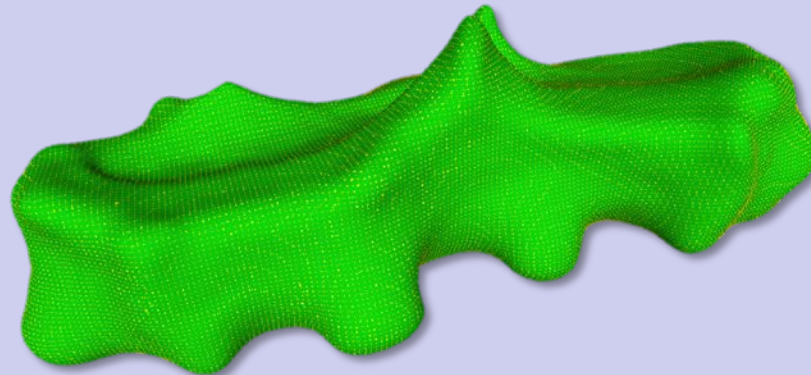
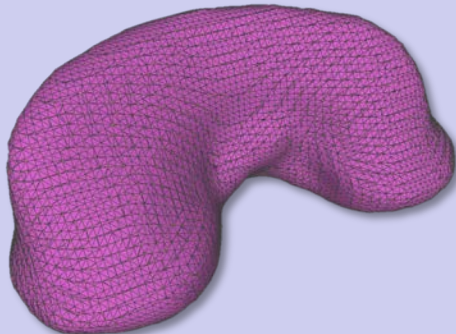
- 2,325 quantified 3-D GSSCP shapes from all 12 subfamilies for morphometric analysis and phylogenetic mapping

## *Anomochloa marantoidea*

Inter-costal  
crescentic

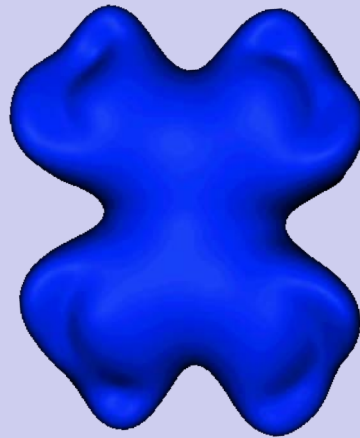
Inter-costal  
polylobate

Costal rondel



# 3-D data workflow: *Outcomes*

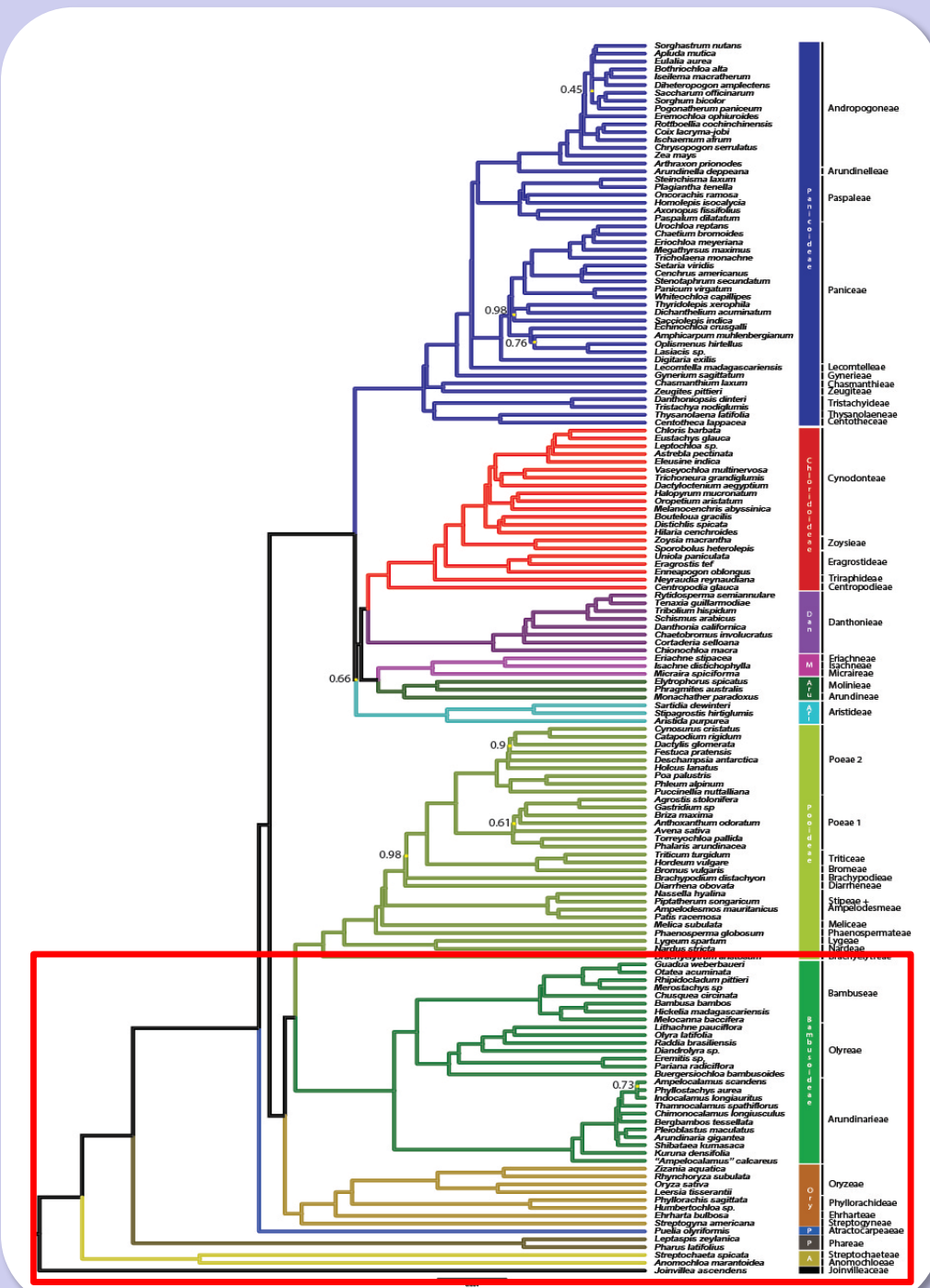
- 2,325 quantified 3-D GSSCP shapes from all 12 subfamilies for morphometric analysis and phylogenetic mapping
- Animations and 3-D printable objects



# Results

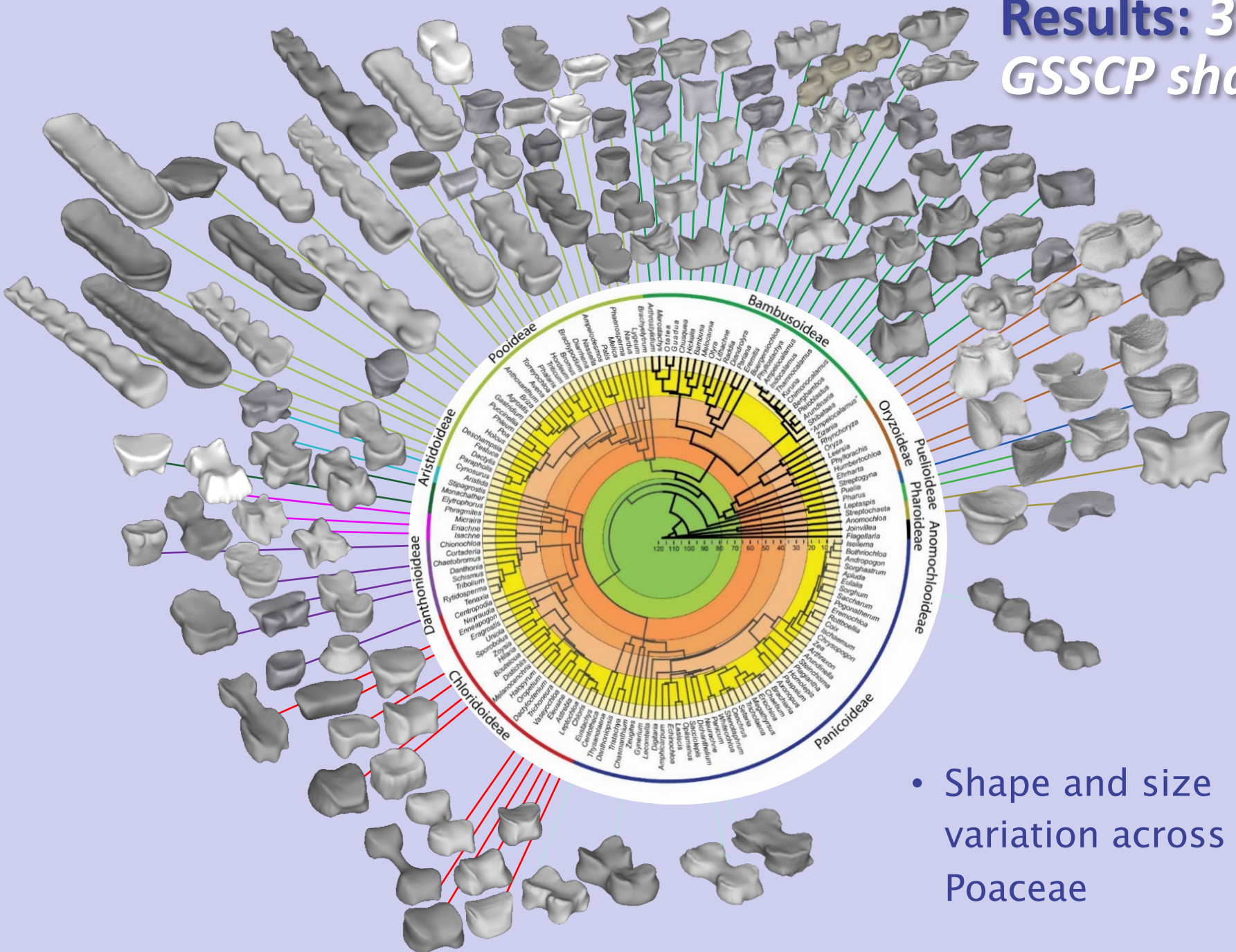
- Focus on early-diverging grasses (A+P+P), Bambusoideae, and Oryzoideae

(Gallaher et al. in prep.)



(Gallaher et al. 2019)

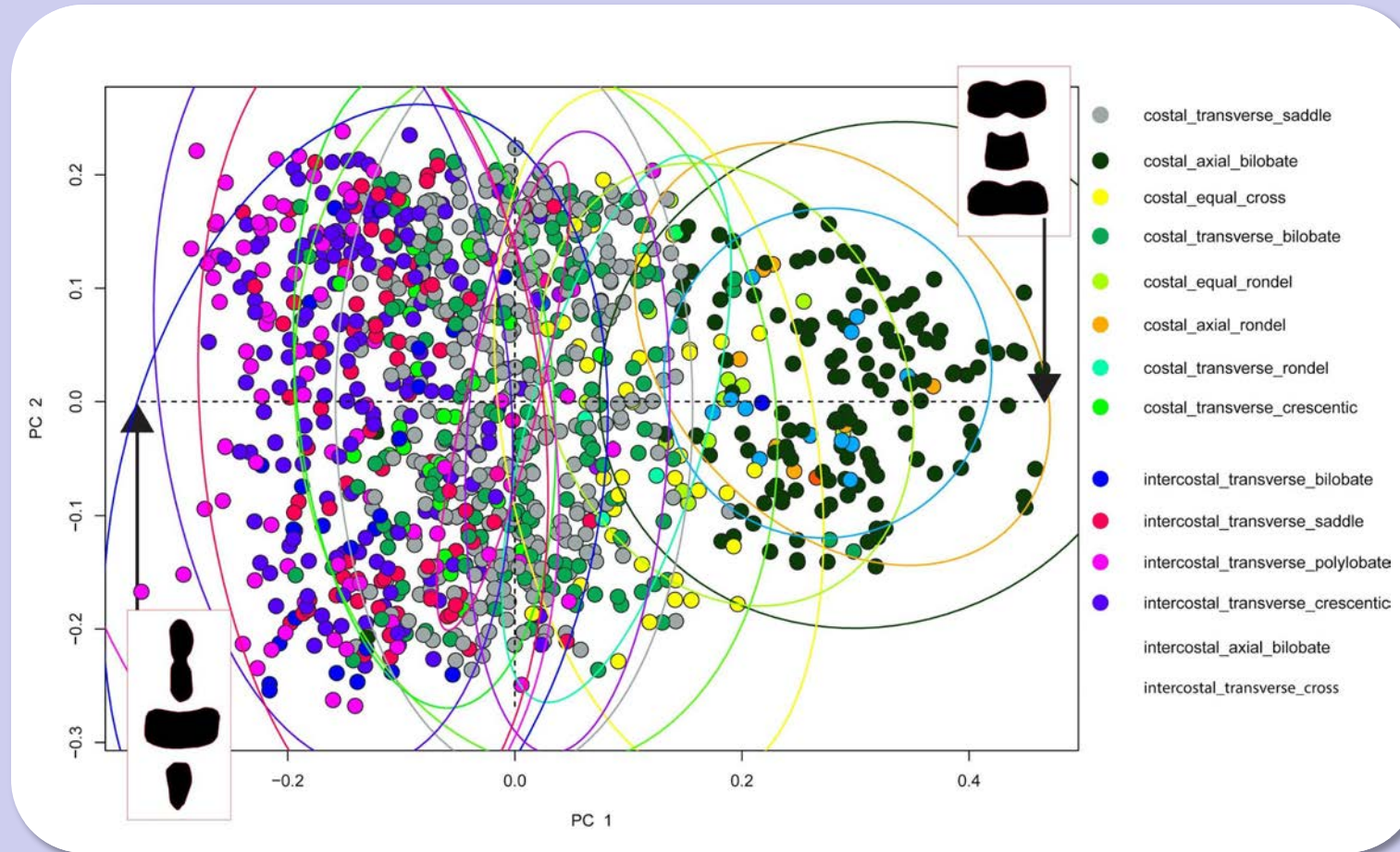
# Results: 3-D GSSCP shape



- Shape and size variation across Poaceae

# Results: 3-D GSSCP shape classification

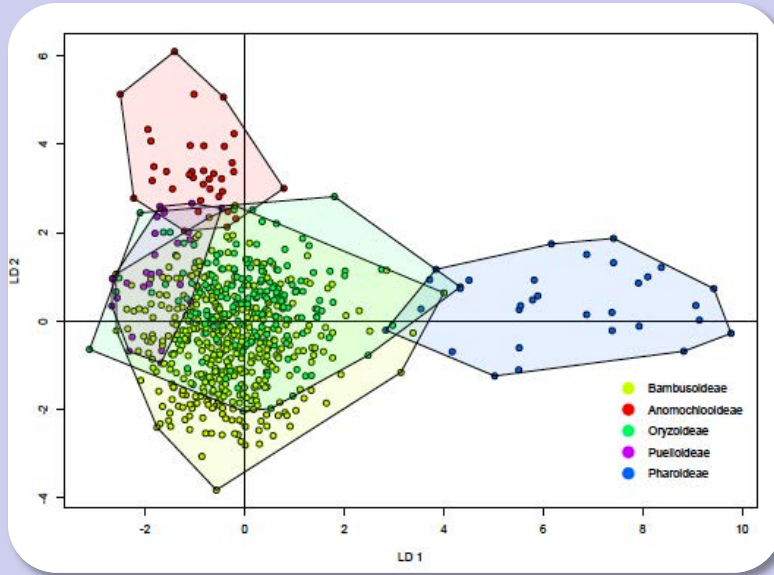
- Principal Component visualization of GSSCP 3-D shape after Procrustes alignment (1,093 images)—there is signal but a lot of noise!



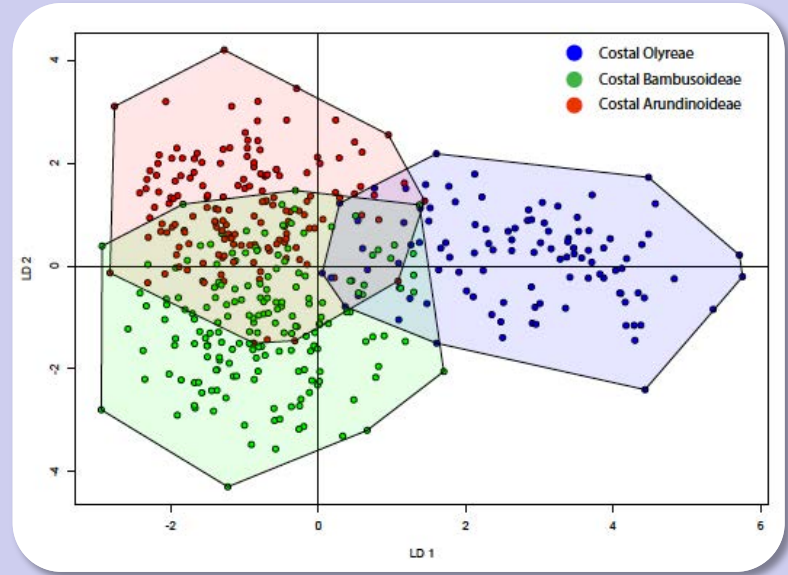
# Results: 3-D GSSCP shape classification

- Linear Discriminant Analysis (LDA) of GSSCP 3-D shape

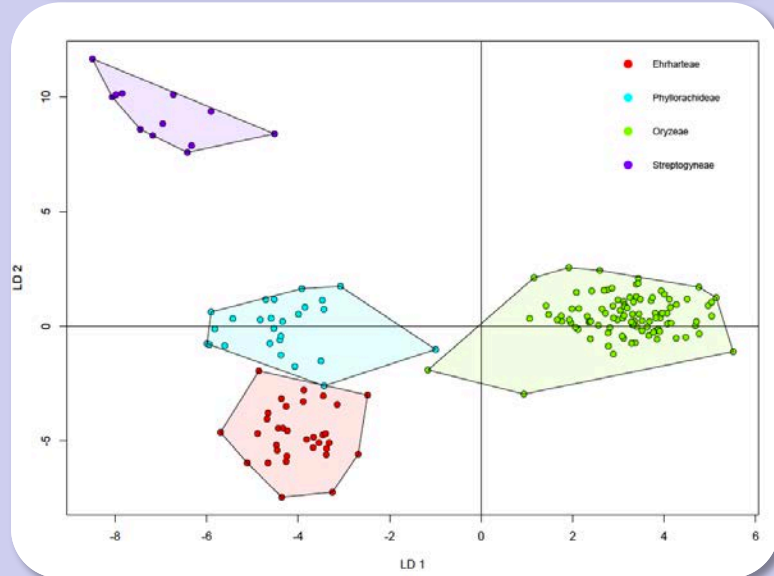
Subfamilies



Bambusoideae tribes



Oryzoideae tribes



- Data = first 66 PCs plus volume, surface area, centroid size, length, width, and height



# Results: 3-D GSSCP shape classification

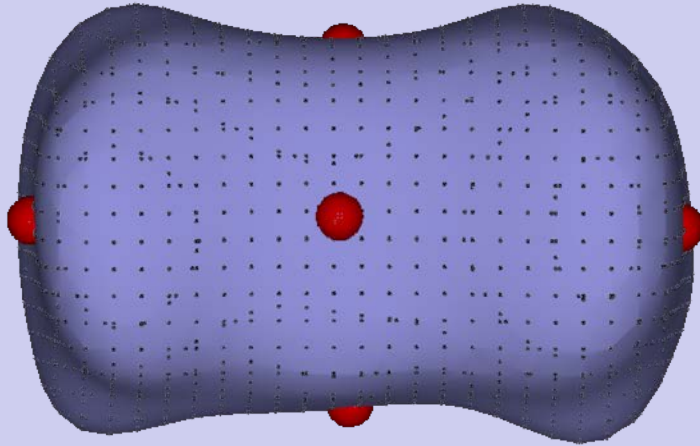
- LDA classification rates (Accuracy) assessed

| Taxon (# classes) | Accuracy LDA: Mean (min-max) | Kappa: Mean (min-max)    | x random chance |
|-------------------|------------------------------|--------------------------|-----------------|
| Subfamily (5)     | 0.7821 (0.6585 - 0.8780)     | 0.5513 (0.3131 - 0.7509) | <b>2.8</b>      |
| Tribe (12)        | 0.6516 (0.5116 - 0.7975)     | 0.5781 (0.4115 - 0.7511) | <b>6.9</b>      |
| Subtribe (22)     | 0.5546 (0.3924 - 0.7381)     | 0.5134 (0.3279 - 0.7141) | <b>11.3</b>     |
| Genus (46)        | 0.4514 (0.3415 - 0.6098)     | 0.4388 (0.3259 - 0.6005) | <b>20.2</b>     |

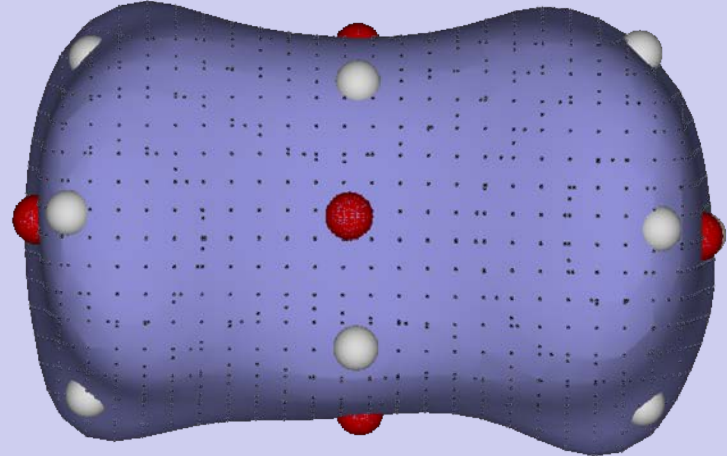
- Correct classification rates are high (3-20x random chance)

# Work in progress

- Switching from 6 to 16 (type 3) landmarks to better fit semi-landmarks to shape



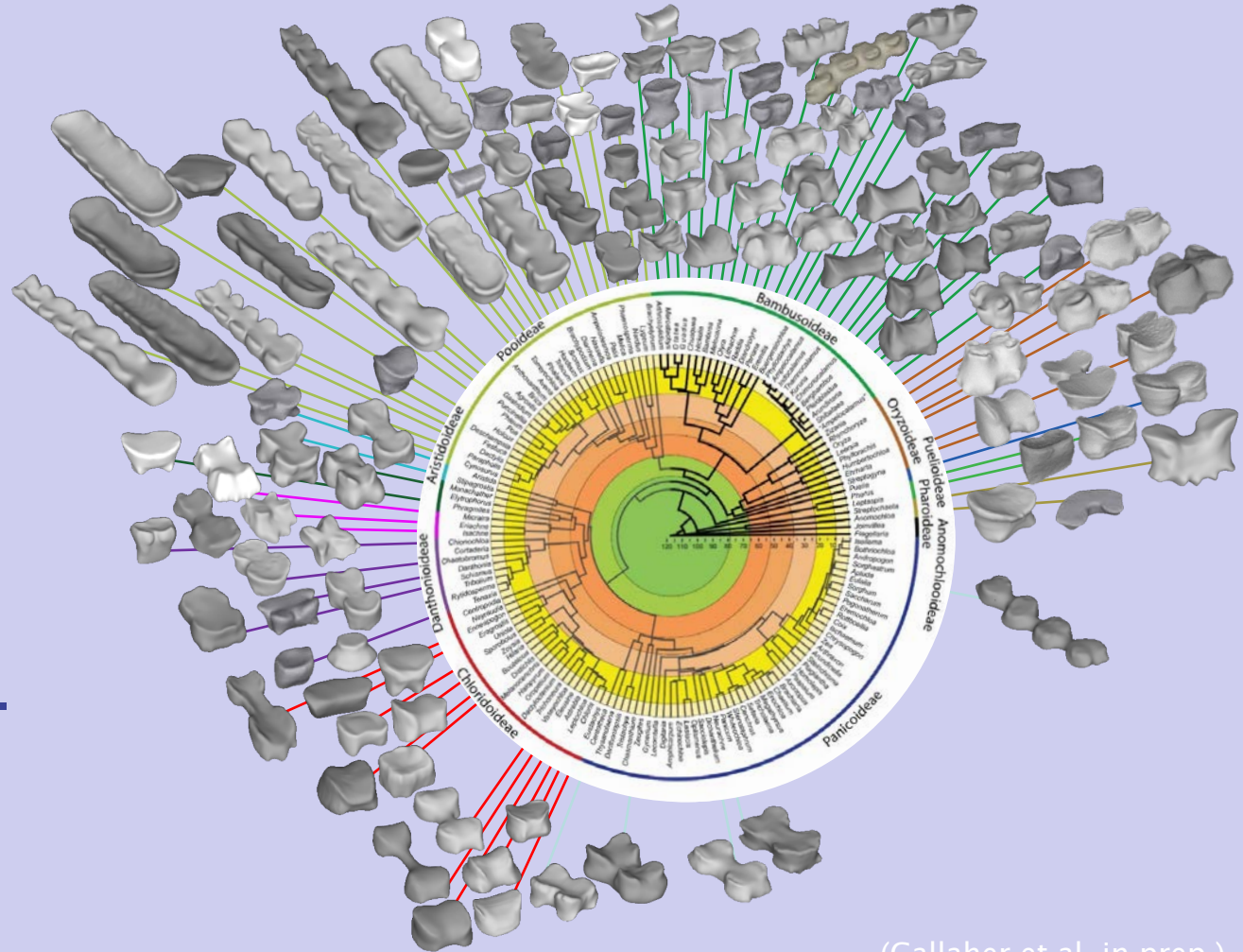
6 (type 3) landmarks



16 (type 3) landmarks

# Work in progress

- Testing for phylogenetic signal in 3-D shape and analyzing shape change across the Poaceae phylogeny

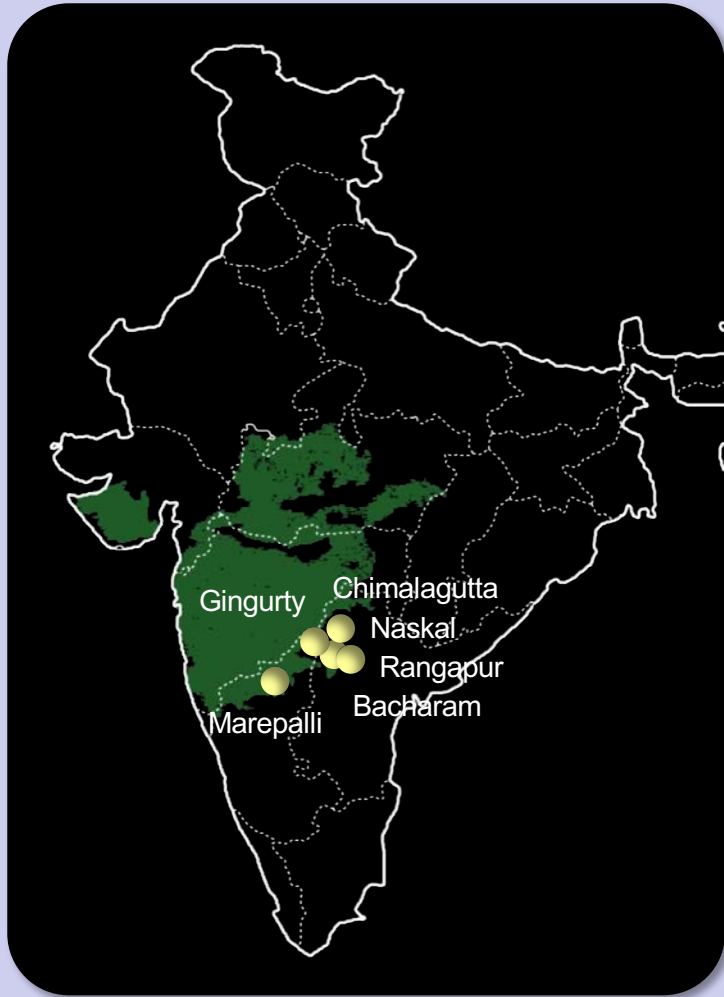


- Mapping grass ecology, physiology etc. and look for correlations

# Creating a phylogenetic key to grass phytolith shape

- Approaches to GSSCP identification/placement
- Properties of GSSCP
- Building a 3-D shape key to GSSCP
- Preliminary application to the Late K-Paleocene of India
- GSSCP and Machine Learning & Computer Vision

# GSSCPs from Late K-Paleocene sites in India



- Study of faunal localities in central India (sedimentology, fauna, flora)
- Late K and Paleocene localities studied so far all have grass phytoliths

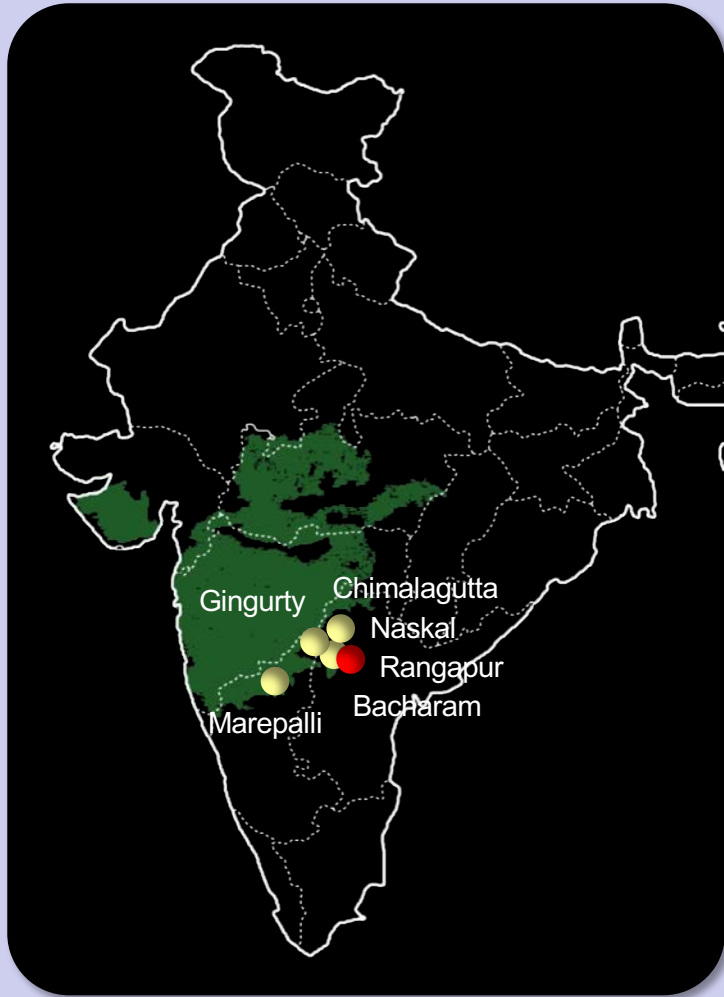


Greg Wilson  
(University of  
Washington)



Jason Moore  
(University of  
New Mexico)

# GSSCPs from Late K-Paleocene sites in India



- Study of faunal localities in central India (sedimentology, fauna, flora)
- Late K and Paleocene localities studied so far all have grass phytoliths
- Preliminary results from the earliest Paleocene(?) Naskal site with abundant grasses

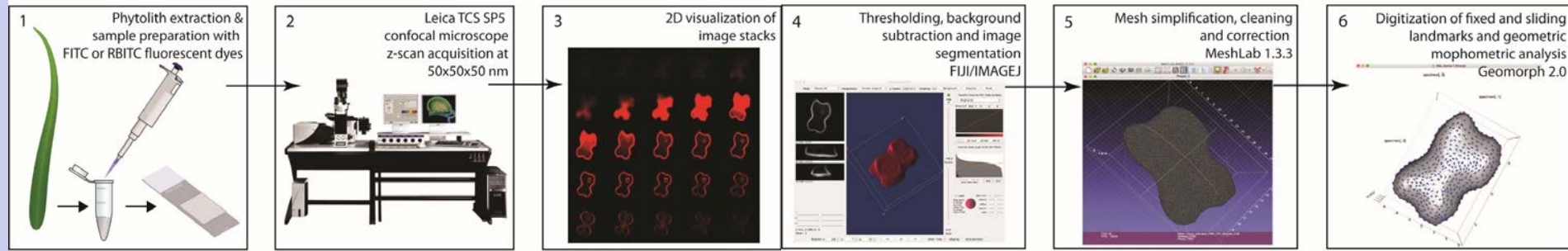


Greg Wilson  
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Washington)



Jason Moore  
(University of  
New Mexico)

# GSSCPs from Naskal (early Paleocene?), India

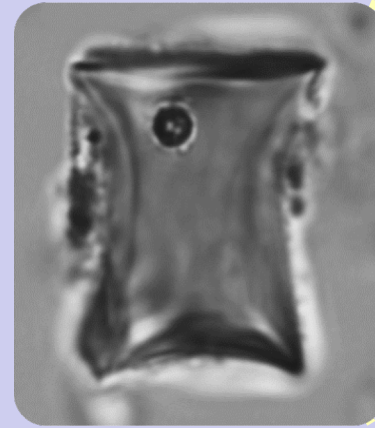


## *3-D data workflow (fossils):*

- Extraction, staining and imaging
- Converting image stacks into 3-D models of GSSCPs
- Fit 6 (type 3) landmarks and 1,000 sliding surface semi-landmark points
- Transform and align meshes using Procrustes superimposition to remove size

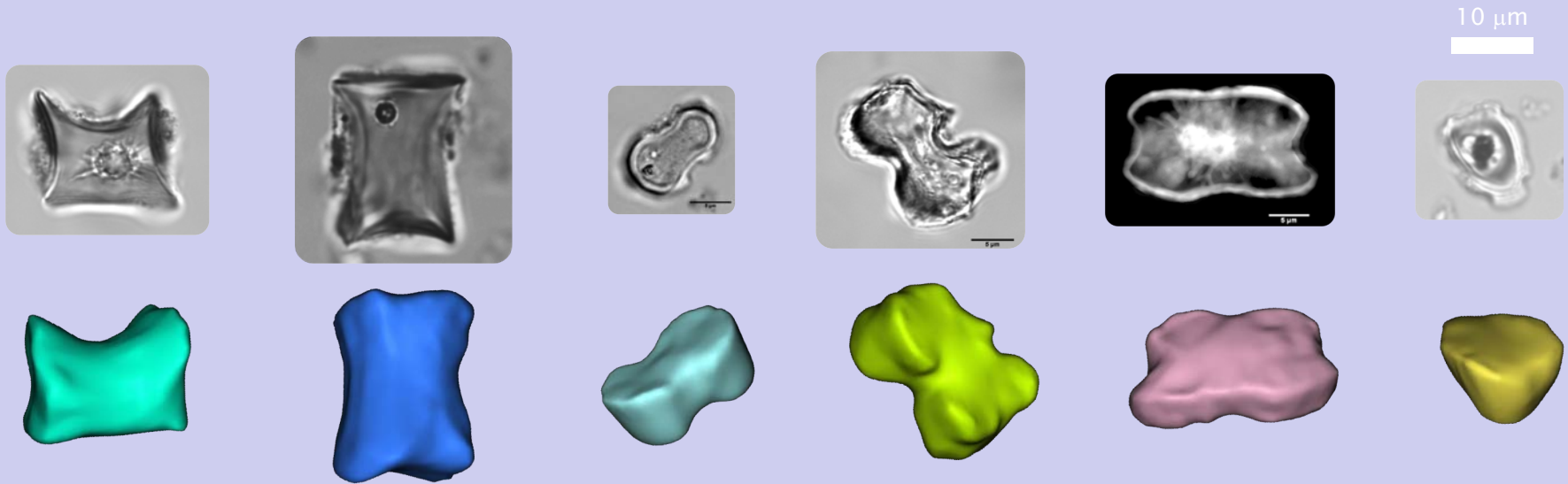
# GSSCPs from Naskal (early Paleocene?), India

- Fossil GSSCPs are covered in debris and are often weathered obscuring fine details
- ... but the the overall shape is visible
- Preliminary results for six GSSCPs





# Taxonomic placements of Paleocene GSSCPs



- Linear Discriminant Analysis (LDA) using our modern dataset to determine taxonomic affinities

# Taxonomic placements of Paleocene GSSCPs



## *Caveats with taxonomic assignment:*

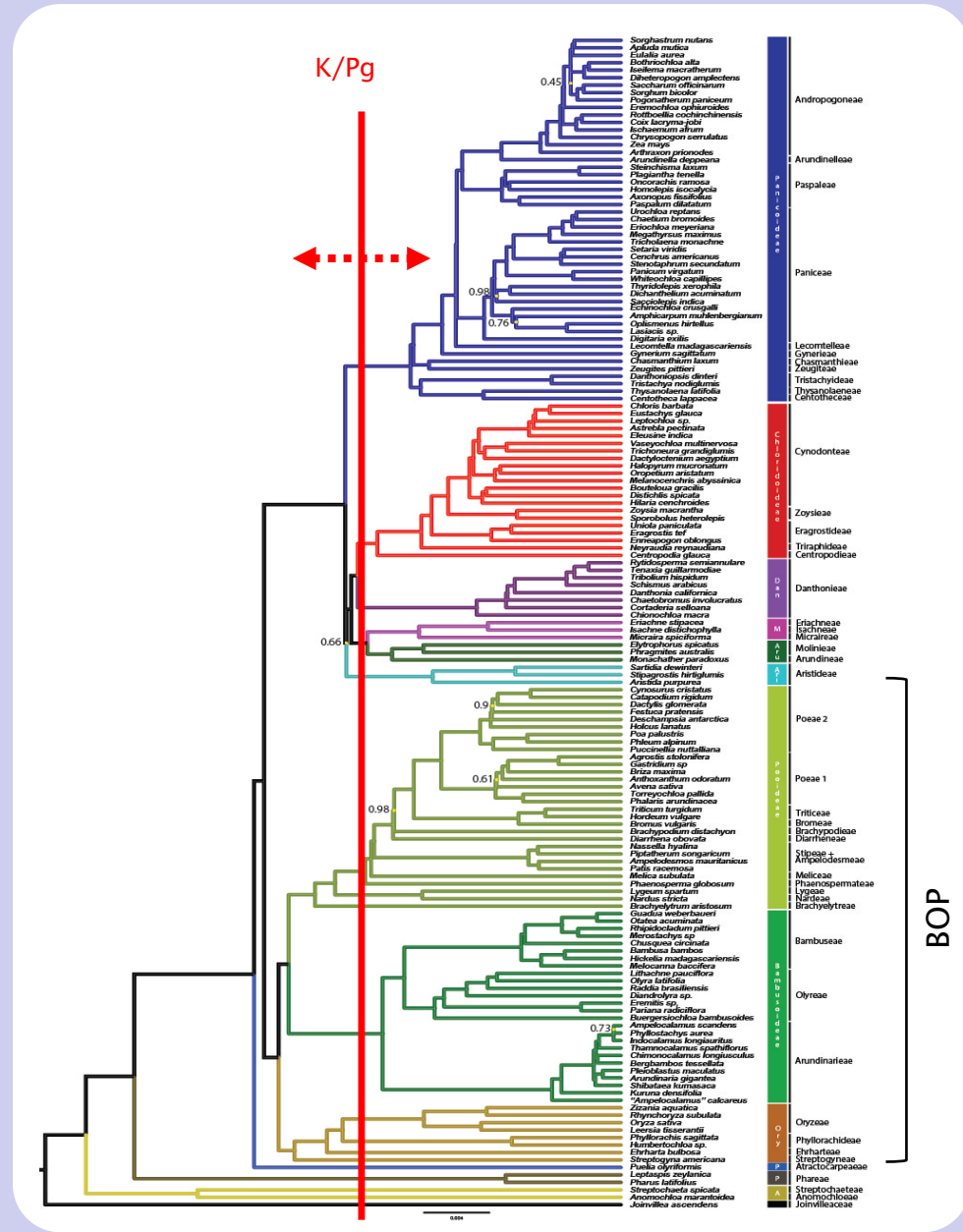
- So far very few GSSCPs have been analyzed
- Modern dataset have only 320 PACMAD specimens
- Phylogeny not explicitly included

# Implications of analysis of Paleocene C/SCPs

- Representatives of BOP clades—similar to previous analyses

Prasad et al. (2005, 2011)

- Supports diversification of BOP well before 66 Ma



# Creating a phylogenetic key to grass phytolith shape

- Approaches to GSSCP identification/placement
- Properties of GSSCP
- Building a 3-D shape key to GSSCP
- Preliminary application to the Late K-Paleocene of India
- GSSCP and Machine Learning & Computer Vision

# Approaches to GSSCP identification/ placement

1. Landmark based 3-D surface morphometrics  
+ 3-D shape  
- landmarks, time consuming
2. Machine Learning and Computer Vision  
+ regular brightfield images, no landmarks  
needed  
- 2-D shape, “black box”



Tim Gallaher



Jessica Perry



Kevin Jamieson

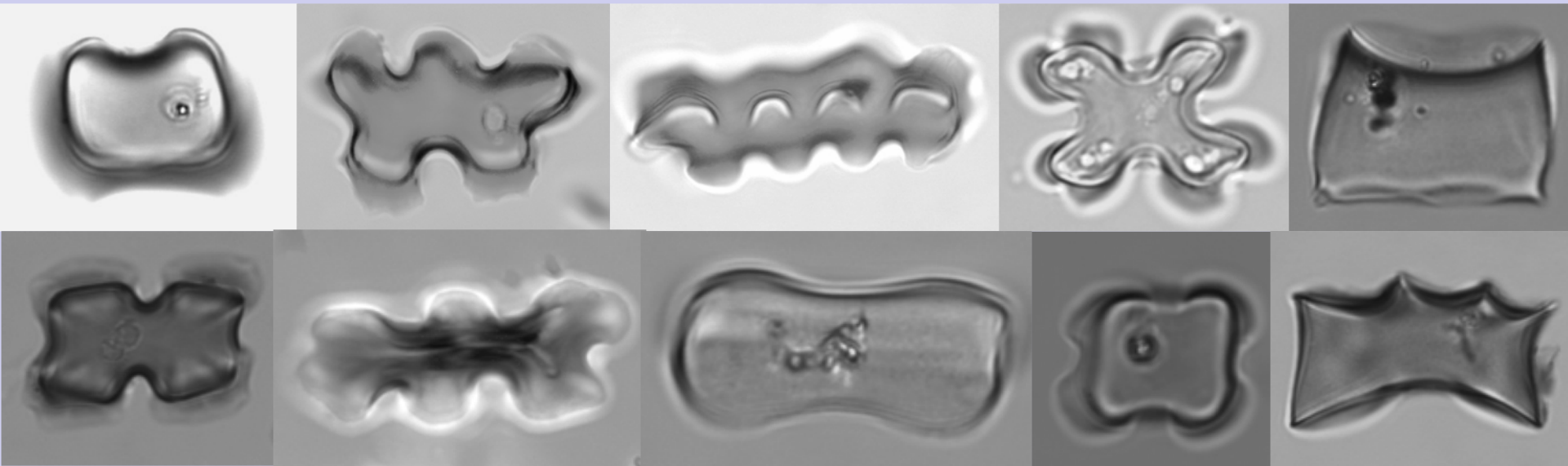
**Other students:**  
Jifan Zhang, Xiangyun  
Meng, Claire Grant

RRF



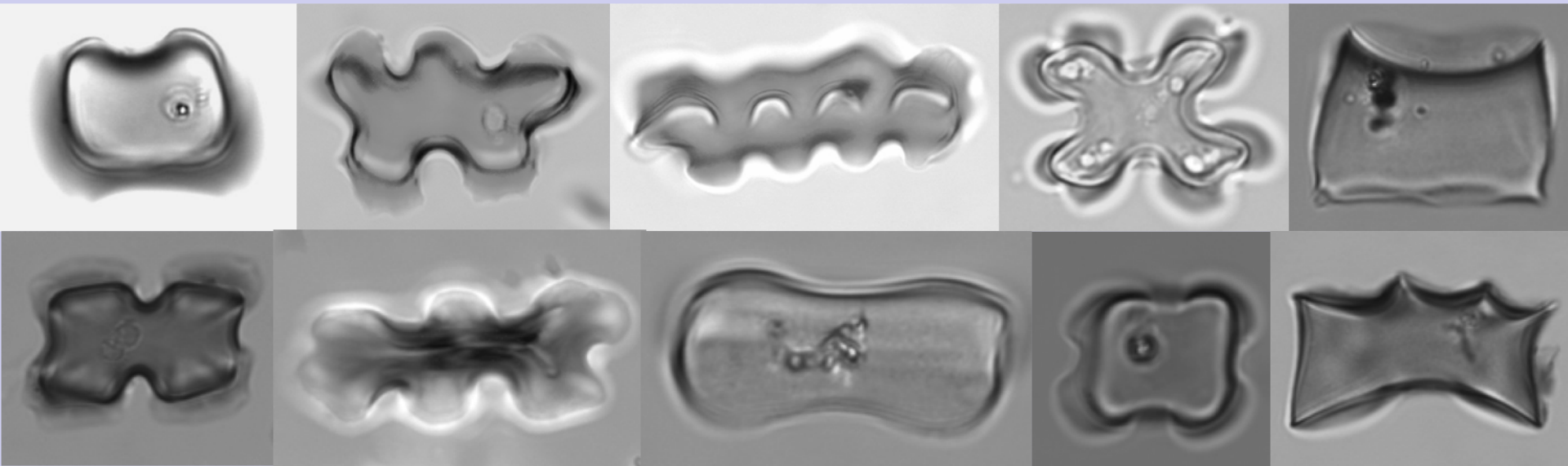
# Machine Learning & Computer Vision (ML&CV)

- Uses training set of 2,480 2-D brightfield images of GSSCP from modern grasses
- Neural network trained to separate taxonomic groups using features of the images
- No alignment or landmarks needed



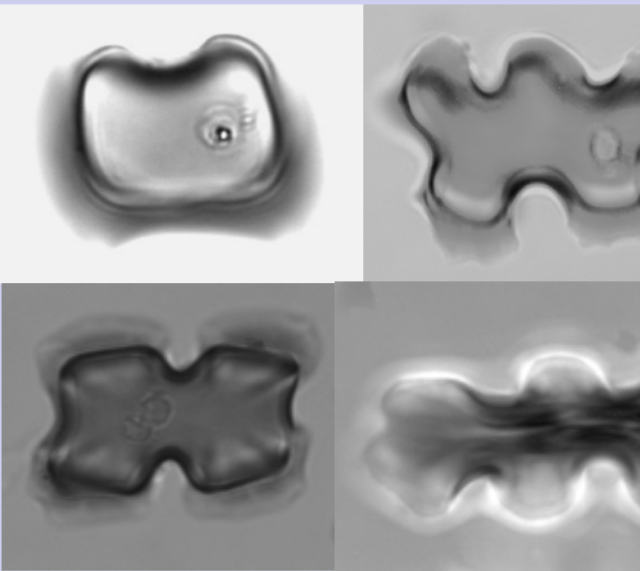
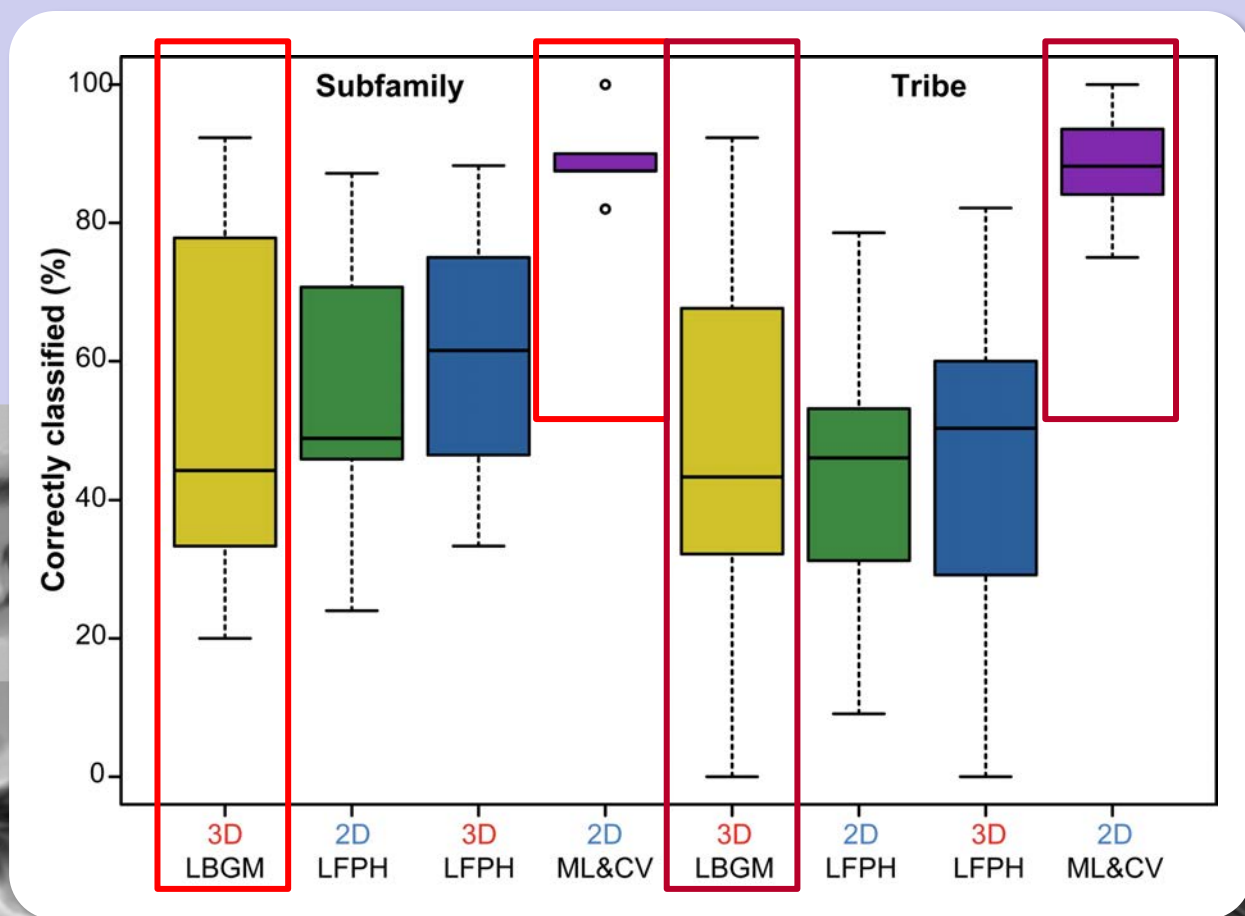
# Machine Learning & Computer Vision (ML&CV)

- Image traits cannot easily be extracted (“black box”)
- Not currently possible to incorporate phylogeny



# Machine Learning & Computer Vision (ML&CV)

- Separation into subfamily and tribe superior to other (2-D and 3-D) methods
- Working to add statistic akin to confidence interval

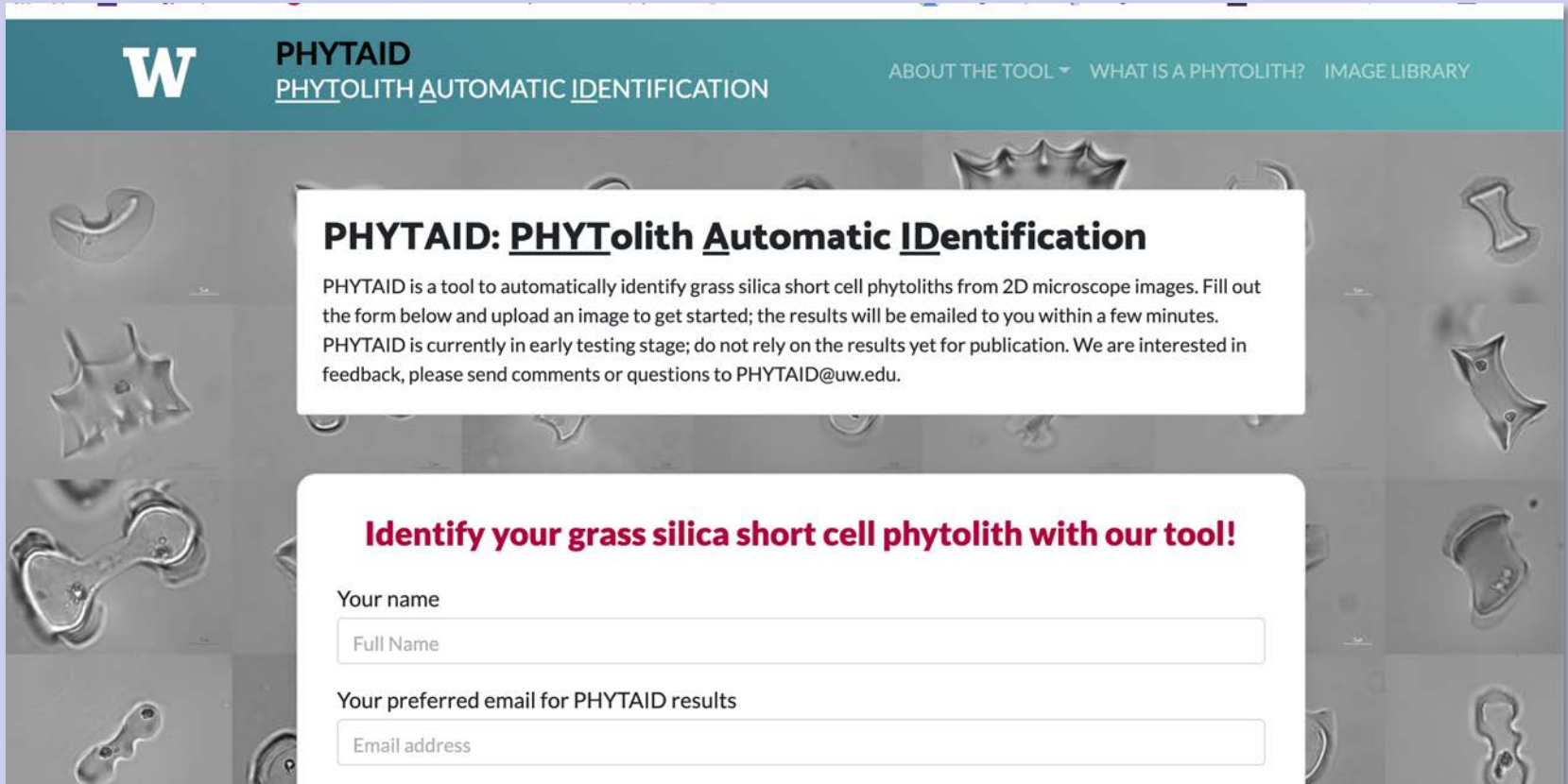




# Sharing data

## PHYTAID (*under construction*):

- Online platform where people can upload images and use ML&CV tool
- ... and download (2-D, 3-D) images, videos, printable models etc.



The screenshot shows the PHYTAID website interface. At the top, there is a teal header with the University of Washington 'W' logo on the left, the text 'PHYTAID PHYTOLITH AUTOMATIC IDENTIFICATION' in the center, and navigation links 'ABOUT THE TOOL', 'WHAT IS A PHYTOLITH?', and 'IMAGE LIBRARY' on the right. Below the header is a grid of microscope images of grass silica short cell phytoliths. A central white box contains the title 'PHYTAID: PHYTolith Automatic IDentification' and a paragraph of text: 'PHYTAID is a tool to automatically identify grass silica short cell phytoliths from 2D microscope images. Fill out the form below and upload an image to get started; the results will be emailed to you within a few minutes. PHYTAID is currently in early testing stage; do not rely on the results yet for publication. We are interested in feedback, please send comments or questions to PHYTAID@uw.edu.' Below this box is another white box with a red heading 'Identify your grass silica short cell phytolith with our tool!' and two input fields: 'Your name' with a 'Full Name' placeholder, and 'Your preferred email for PHYTAID results' with an 'Email address' placeholder.

**PHYTAID: PHYTolith Automatic IDentification**

PHYTAID is a tool to automatically identify grass silica short cell phytoliths from 2D microscope images. Fill out the form below and upload an image to get started; the results will be emailed to you within a few minutes. PHYTAID is currently in early testing stage; do not rely on the results yet for publication. We are interested in feedback, please send comments or questions to PHYTAID@uw.edu.

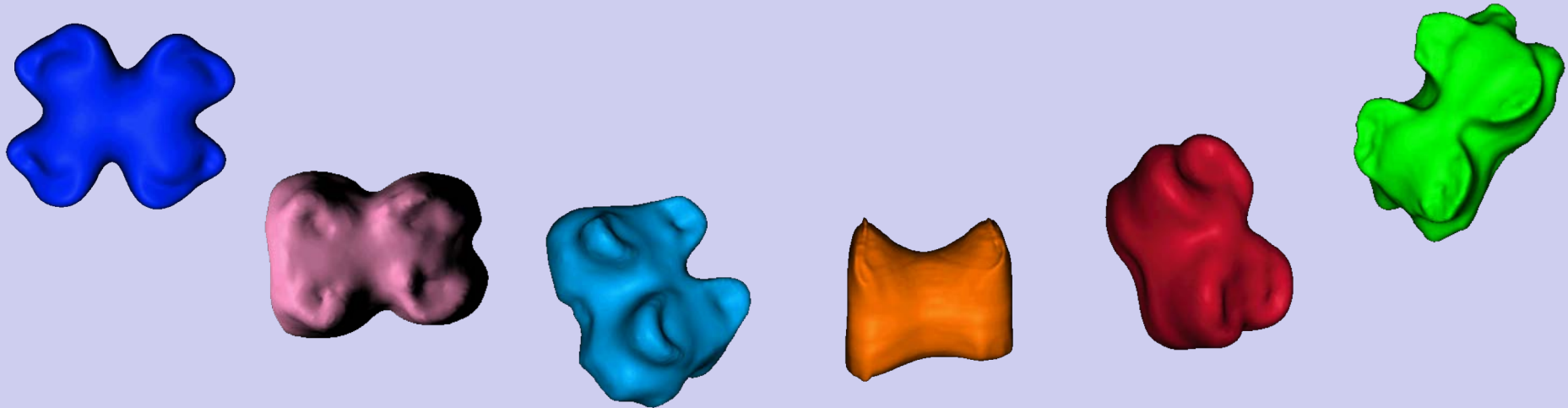
**Identify your grass silica short cell phytolith with our tool!**

Your name  
Full Name

Your preferred email for PHYTAID results  
Email address

# Conclusions

- Quantitative methods (e.g., 3-D morphometrics, machine learning) have a great potential for making taxonomic placement of grass phytoliths more precise and robust
- Ultimately this will enhance research on grasses in paleobotany, paleoecology and archaeobotany



<http://www.burkemuseum.org/phytaid>

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Thank you for your  
attention!



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