

# Introduction to statistical shape analysis II: Semi- landmarks and beyond

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2/19/20



# Outline

- How many landmarks are enough?
- Semi-landmarks
- Deformable Analysis
- Spherical Harmonic Representation



# How many landmarks are enough to characterize shape and size variation?

- Criteria for landmarks to be homologous and reproducible results in very sparse data from images
- For many data sets, this may not be sufficient to capture shape changes, especially along curves or smooth surfaces
- Point-landmark too stringent for effective biometrics in many 3D applications

Watanabe, Akinobu. "How many landmarks are enough to characterize shape and size variation?." PloS one 13.6 (2018): e0198341.



# How many landmarks are enough to characterize shape and size variation?

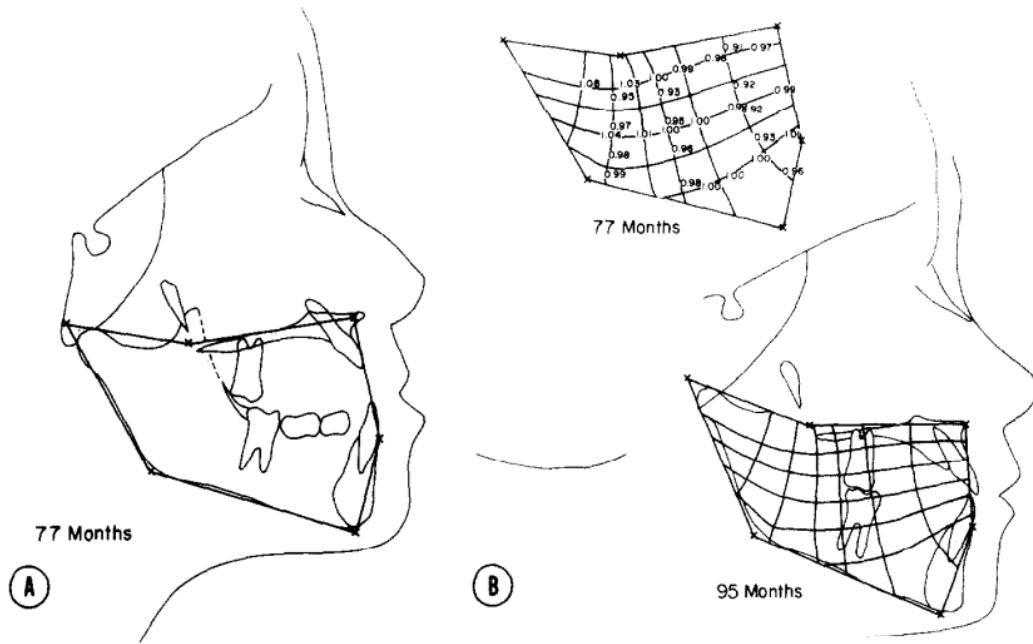
- Criteria for landmarks to be homologous and reproducible results in very sparse data from images
- For many data sets, this may not be sufficient to capture shape changes, especially along curves or smooth surfaces
- Point-landmark too stringent for effective biometrics in many 3D applications

**Image data provides rich phenotype descriptions – how can this be leveraged?**

Watanabe, Akinobu. "How many landmarks are enough to characterize shape and size variation?." PloS one 13.6 (2018): e0198341.



# Constructed landmarks

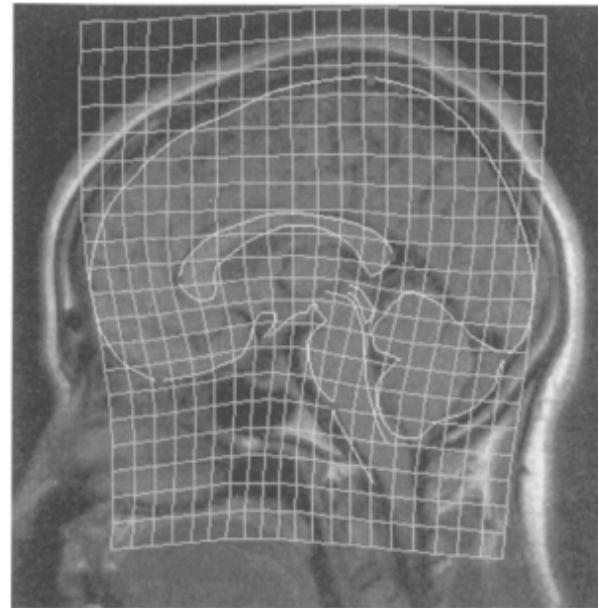
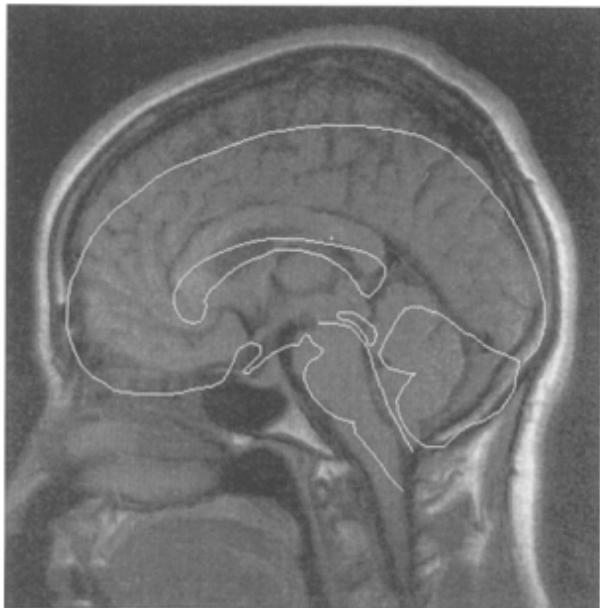


Geometric combinations of defined landmarks along lines erected at specific angles to define new landmarks

Moyers, Robert E., and Fred L. Bookstein. "The inappropriateness of conventional cephalometrics." *American Journal of Orthodontics and Dentofacial Orthopedics* 75.6 (1979): 599-617.



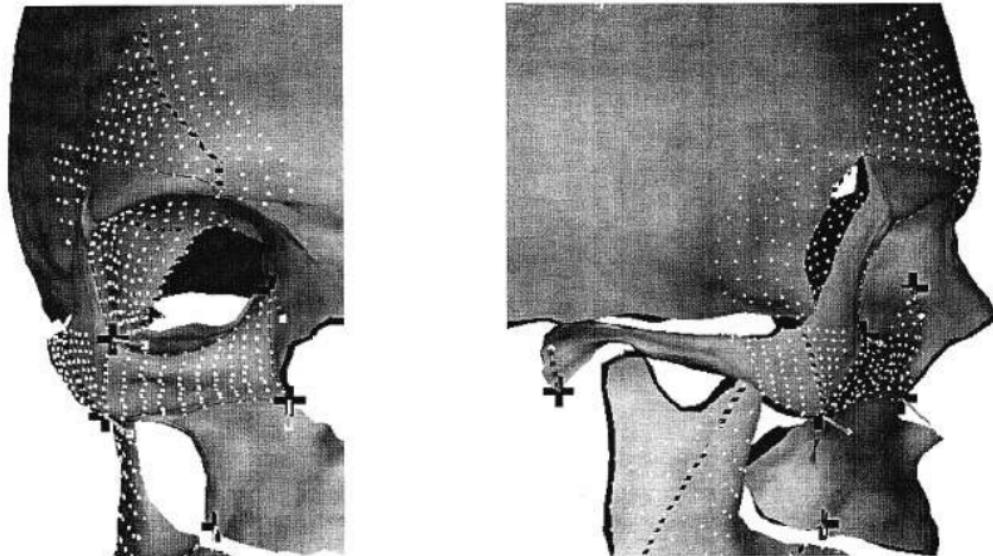
# Extensions of TPS to include curvature



Bookstein, Fred L., and William DK Green. "A feature space for edgels in images with landmarks." *Journal of Mathematical imaging and vision* 3.3 (1993): 231-261.



# Smooth surface analysis

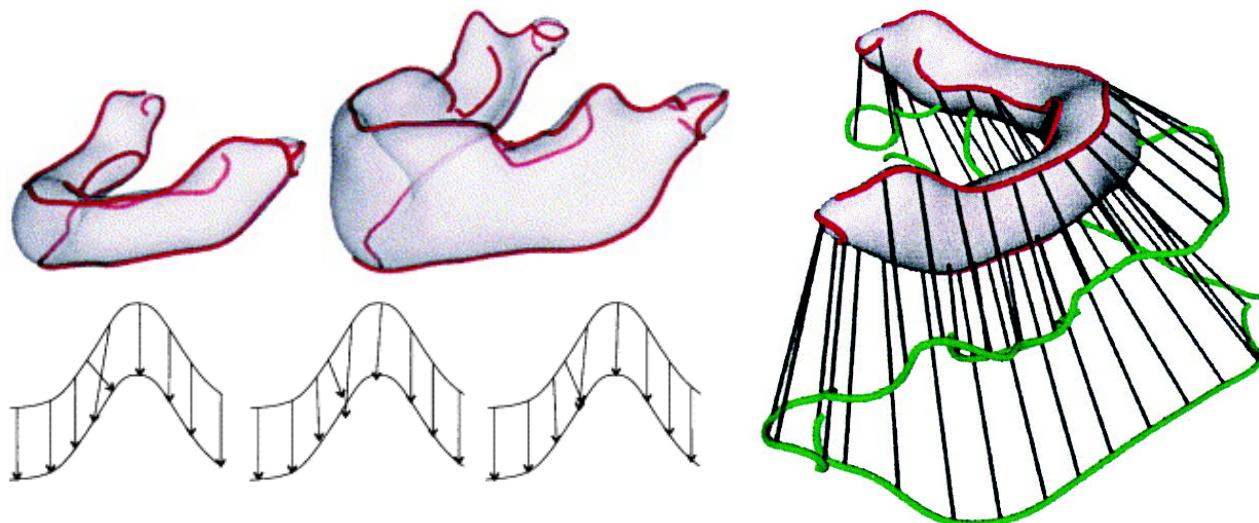


- Thin-plate spline unwarping to the Procrustes mean configuration
- Equally spaced points are declared homologous along curves
- Evenly spaced points are declared homologous on surface patches derived from these

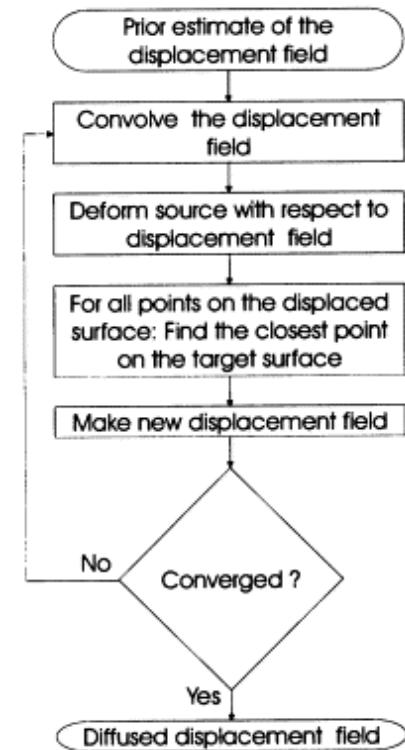
Cutting, Court, et al. "A three-dimensional smooth surface analysis of untreated Crouzon's syndrome in the adult." *The Journal of craniofacial surgery* 6.6 (1995): 444-453.



# Geometry constrained diffusion



- Ridge lines are automatically extracted from surface curvature
- Curves are matched in order to establish object correspondence
- Semilandmarks are mapped into Procrustes space for analysis

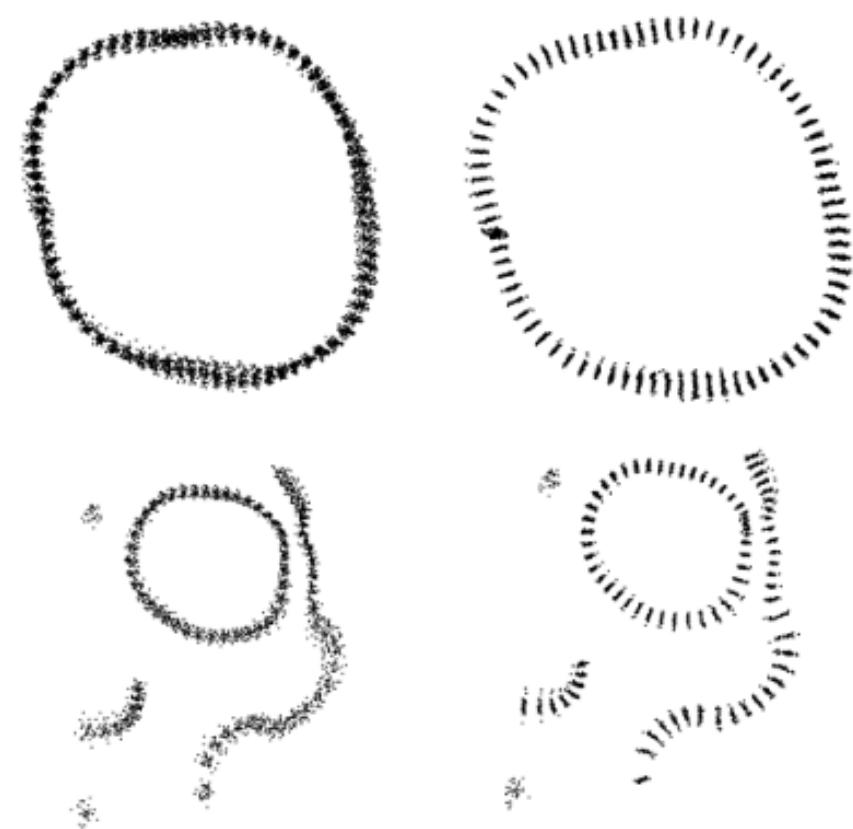


Andresen, Per Rønsholt, and Mads Nielsen. "Non-rigid registration by geometry-constrained diffusion." *Medical Image Analysis* 5.2 (2001): 81-88.



# Sliding semi-landmarks

- Begin with structures that are known to correspond as parts (classical homology)
- Represents them by geometric curves or surfaces that generate reasonable mapping functions
- After Procrustes superimposition, semi-landmarks are slid along the surface to optimize correspondences



Before and after semi-landmark alignment of skull



# What is wrong with equidistant samples?

Produce spacing as a by-product of the analysis since the analysis is ignorant of the actual spacing.



Figure 3. (a) Form with one true landmark in the lower left corner and 31 other points equally spaced along the outline. (b) Bent form with one true landmark (1) and 31 other points in equal spacing. (c) The position of the points now optimizes bending energy.

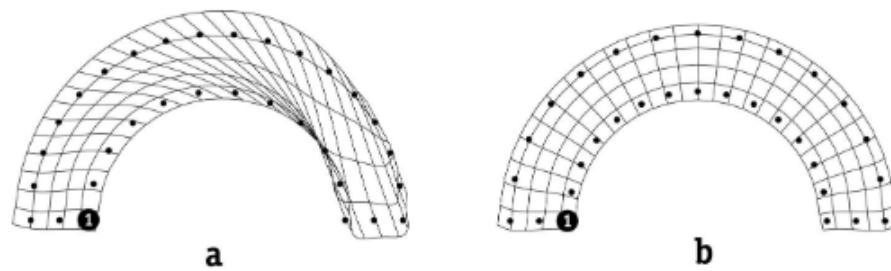


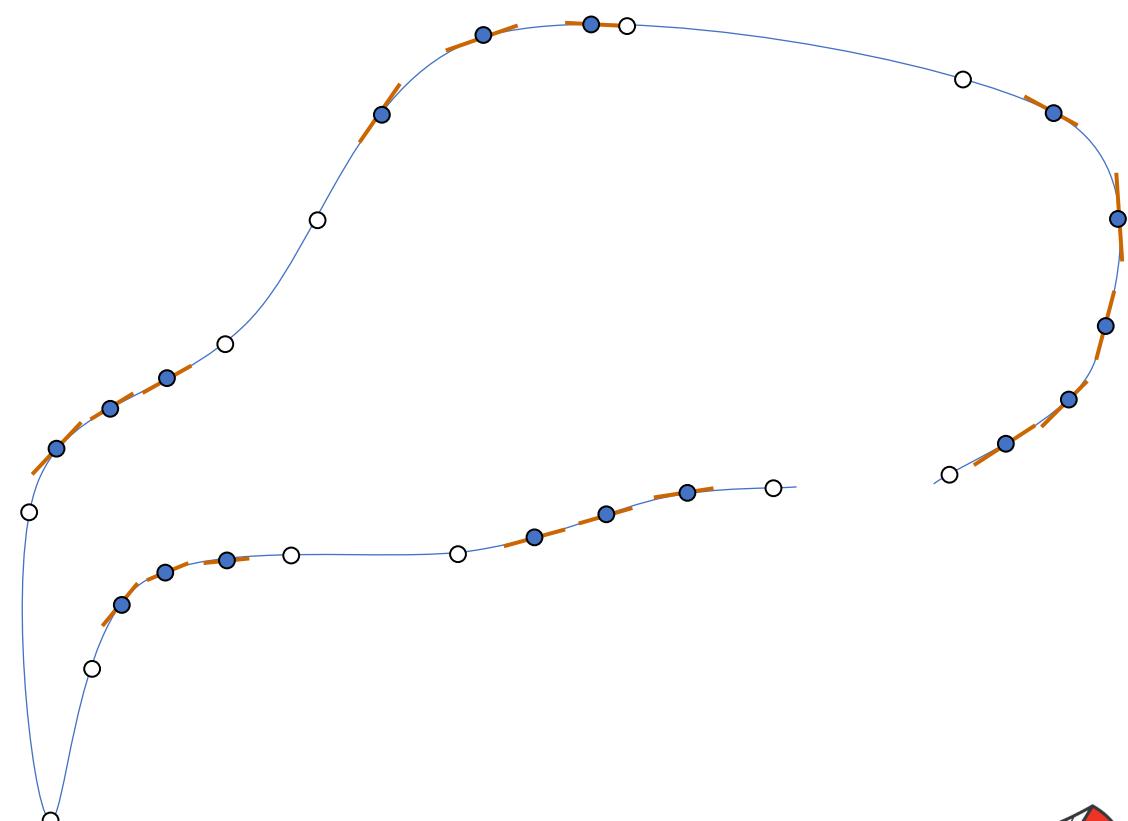
Figure 4. Splines corresponding to Figure 3. (a) Deformation grid from the form in Figures 3a and 3b. (b) Deformation grid from the form in Figures 3a and 3c.

Gunz, Philipp, Philipp Mitteroecker, and Fred L. Bookstein. "Semilandmarks in three dimensions." *Modern morphometrics in physical anthropology*. Springer, Boston, MA, 2005. 73-98.



# Sliding semi-landmark method

- 1) Find optimal Procrustes alignment of samples using landmark points
- 2) Slide semi-landmark points along the surface until they satisfy matching criteria with a single reference specimen
- 3) Calculate Procrustes average shape
- 4) Slide semi-landmark points along the surface until they satisfy matching criteria with the Procrustes average
- 5) Repeat steps 3 and 4 until convergence



Bookstein, Fred L. "Landmark methods for forms without landmarks: morphometrics of group differences in outline shape." *Medical image analysis* 1.3 (1997): 225-243.

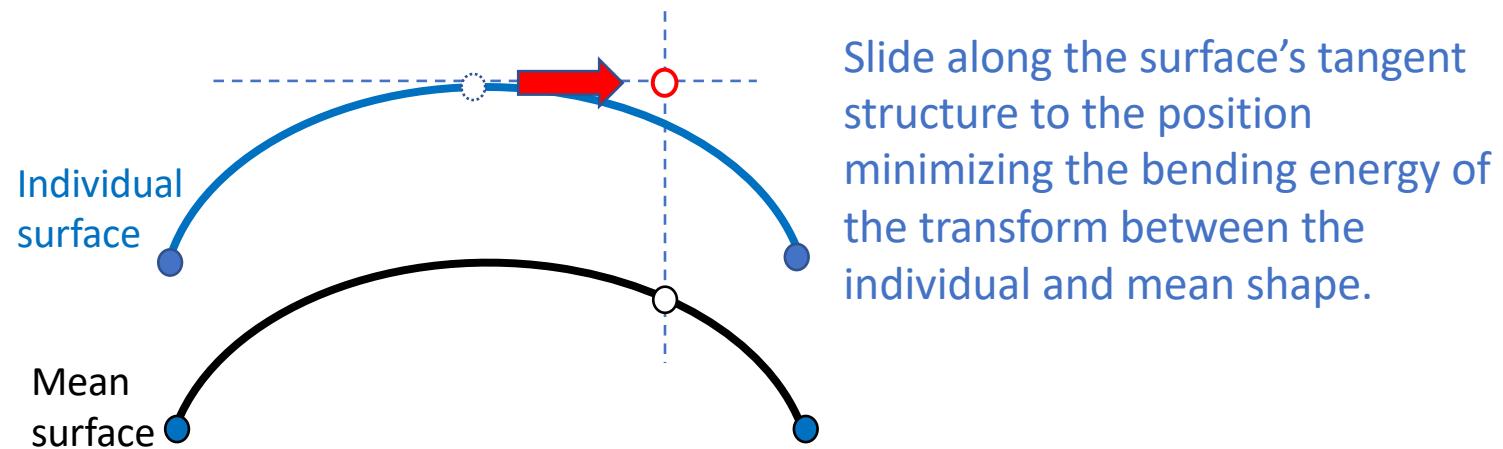


# Determining position of sliding semi-landmarks

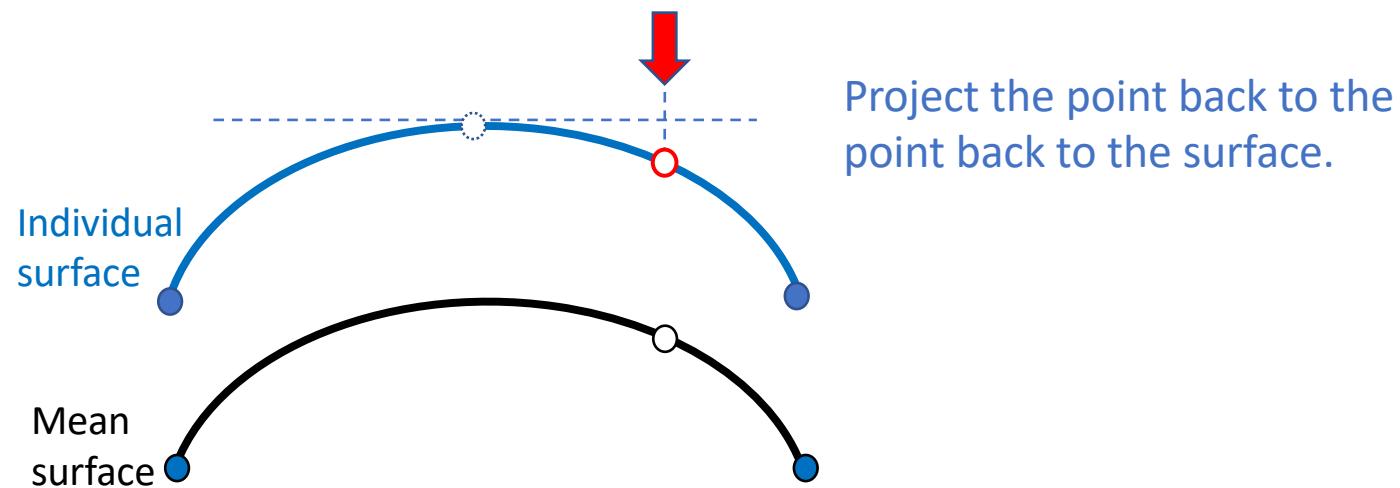
- 1) **Minimum bending energy criterion:** select semi-landmark positions that result in the smoothest possible transformation to the mean shape
- 2) **Procrustes distance criterion:** estimate the tangent to the mean surface for each semi-landmark point and remove the component of the difference between the mean and each specimen that lies along this tangent.



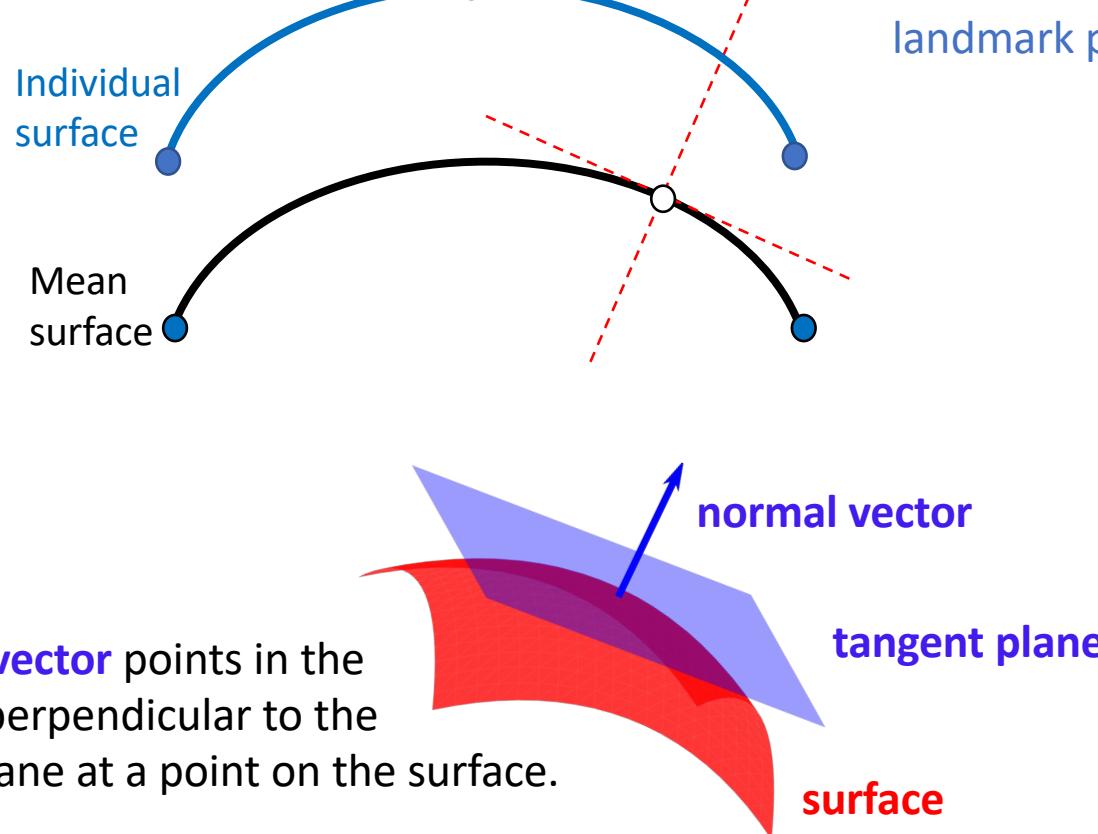
# Minimum bending energy criteria



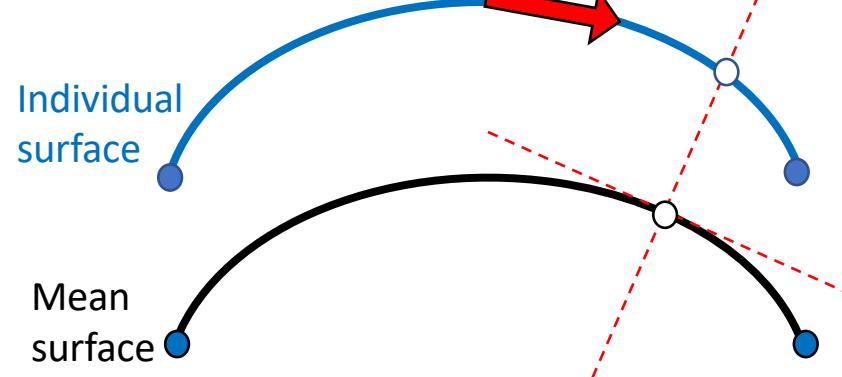
# Minimum bending energy criteria



# Procrustes distance criteria



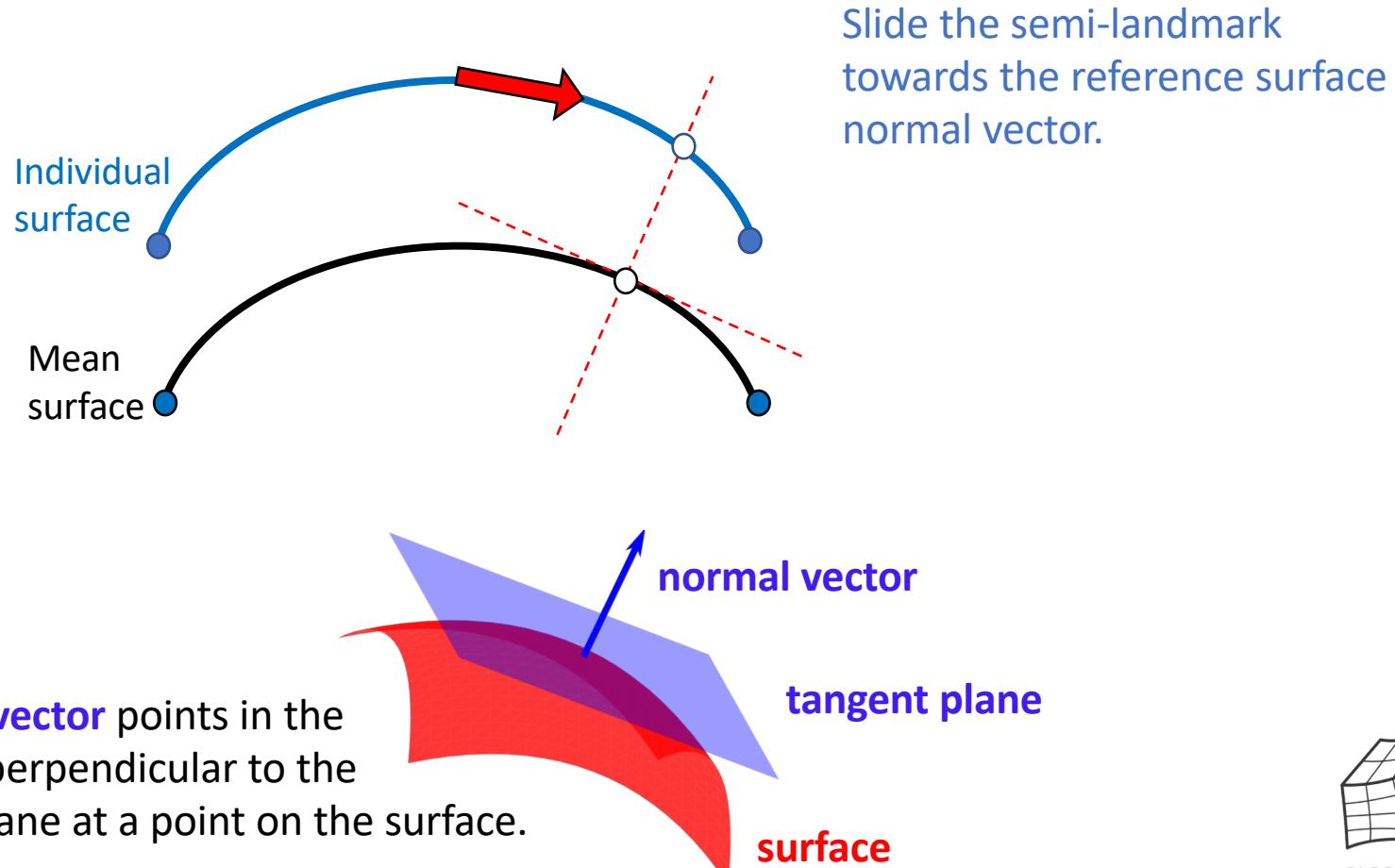
# Procrustes distance criteria



Slide the semi-landmark  
towards the reference surface  
normal vector.



# Procrustes distance criteria



# Minimum bending energy or Procrustes distance?

- Use different background assumptions
- The difference between the criteria can alter the results when morphological variation in the sample is low
- More noticeable with smaller numbers of semi-landmarks

Perez, S. Ivan, Valeria Bernal, and Paula N. Gonzalez. "Differences between sliding semi-landmark methods in geometric morphometrics, with an application to human craniofacial and dental variation." *Journal of anatomy* 208.6 (2006): 769-784.



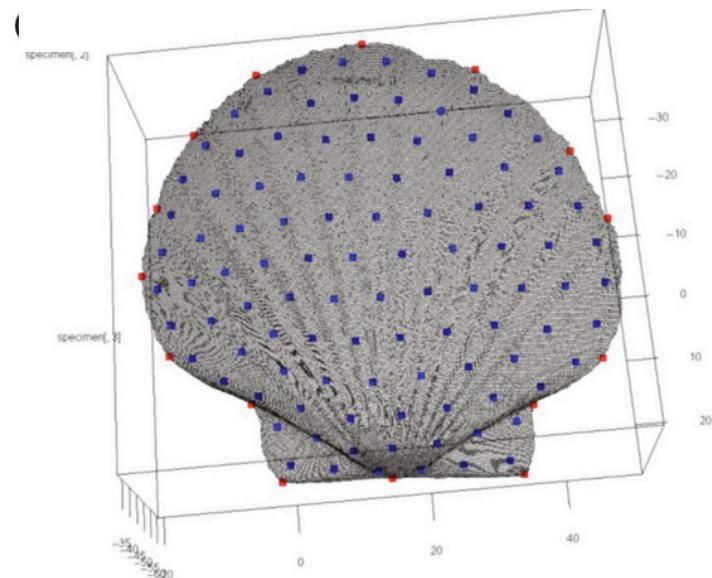
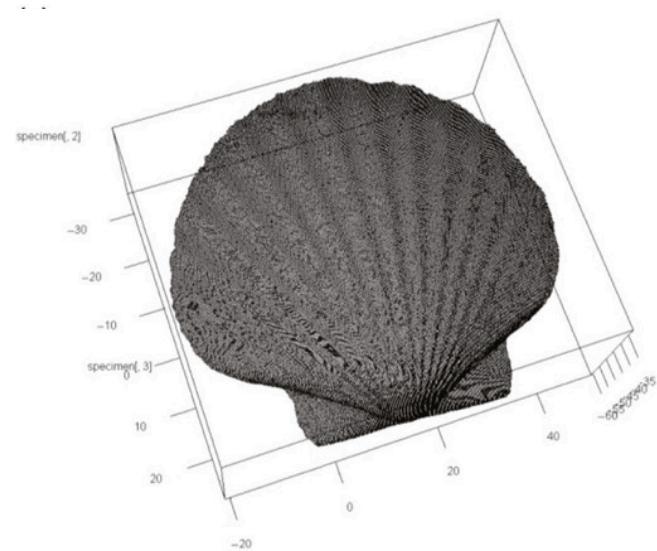
# Limitations of semi-landmarks

- Limited homology
- The method of handling semi-landmarks can influence the results
- The number of semi-landmarks may also influence the results
- Sliding semi-landmark positions are dependent on the dataset. Not possible to compare new shapes without recalculating



# Semi-landmarks in R: Morpho

R Toolboxes [Morpho](#) and [Geomorph](#) for morphometric analysis provide support for capturing and analyzing semi-landmarks



# Coming soon

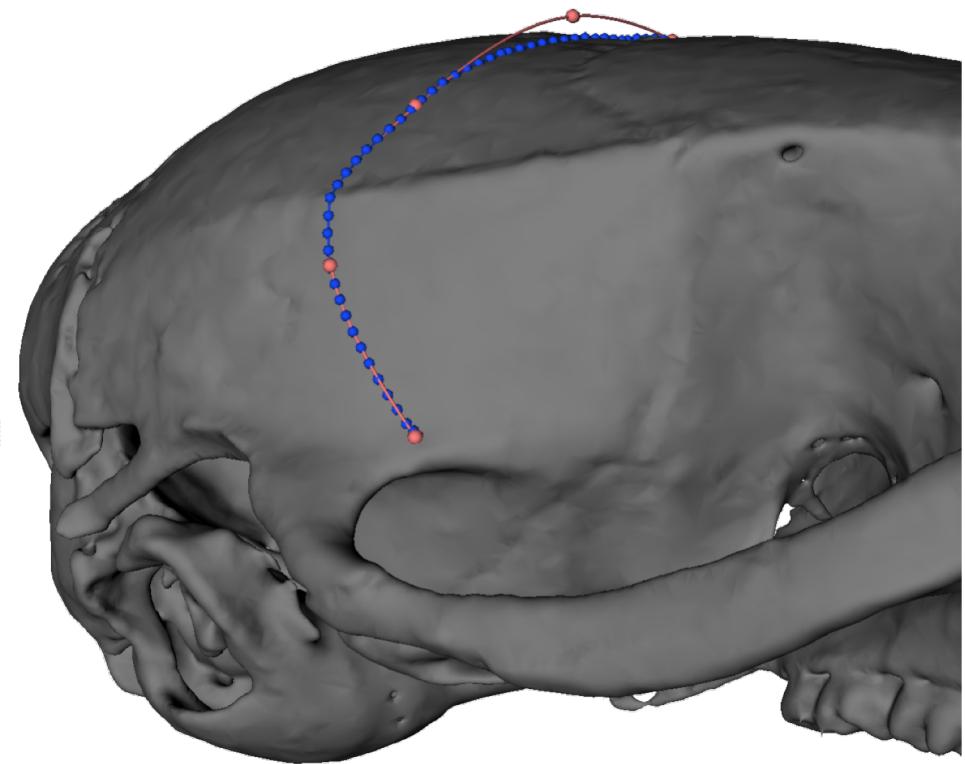
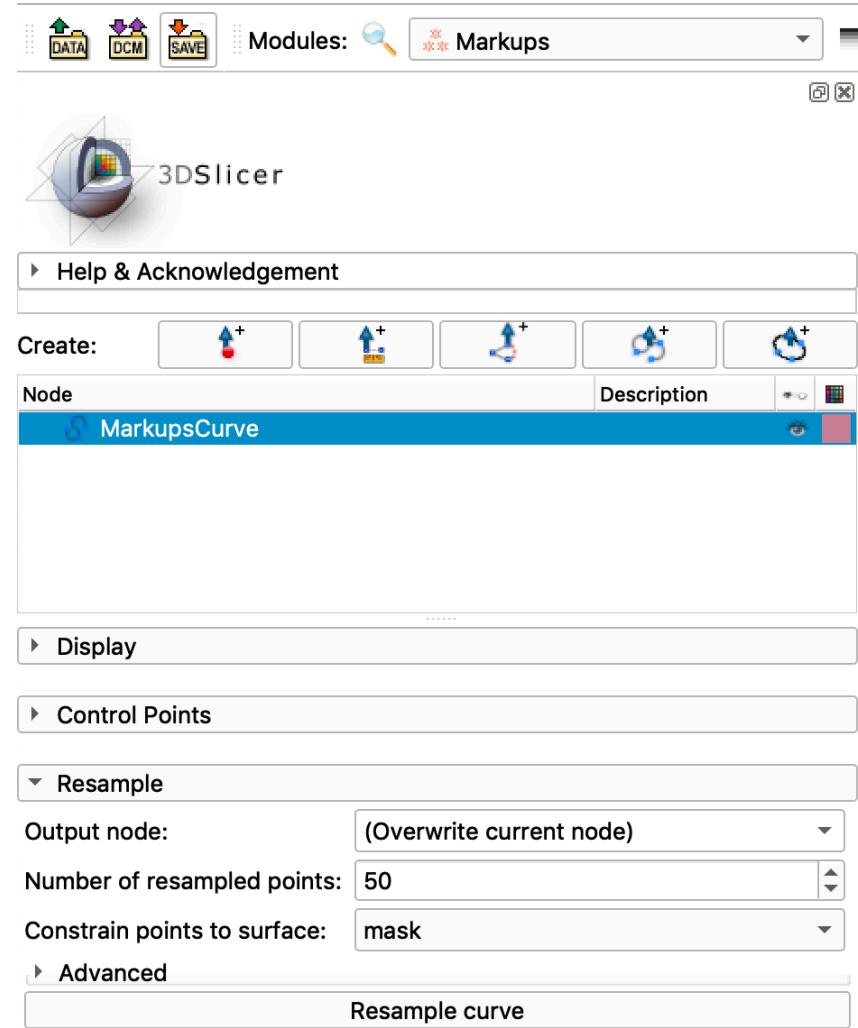
The image shows the 3DSlicer software interface. At the top, there is a toolbar with various icons for file operations (DATA, DCM, SAVE), module selection (Modules: SemiLandmark), and other tools. Below the toolbar is a Python Interactor window displaying the following text:

```
Python 3.6.7 (default, Jul 28 2019, 23:06:36)
[GCC 4.2.1 Compatible Apple LLVM 8.0.0 (clang-800.0.42.1)] on darwin
>>>
Loading Slicer RC file [/Users/srolfe/.slicercrc.py]
/Users/srolfe/Dropbox/SlicerWorkspace/SMwSML/SemiLandmark/Resources/UI/
```

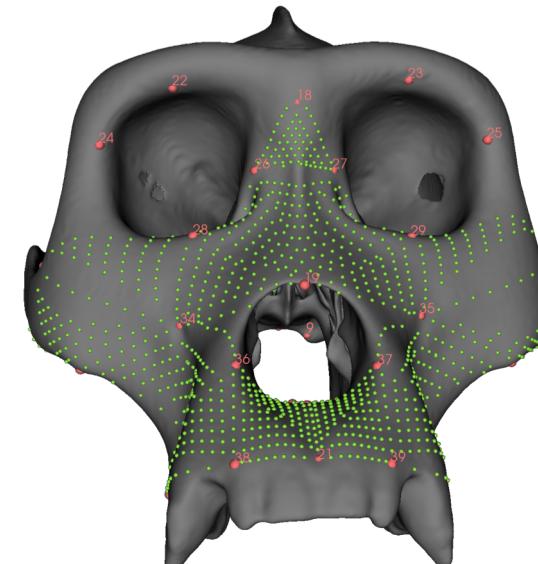
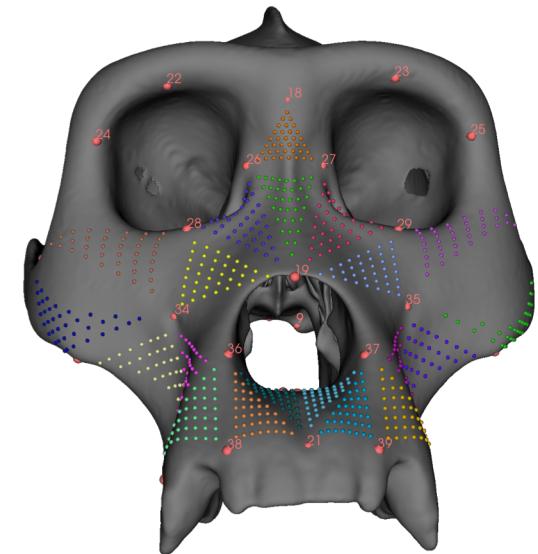
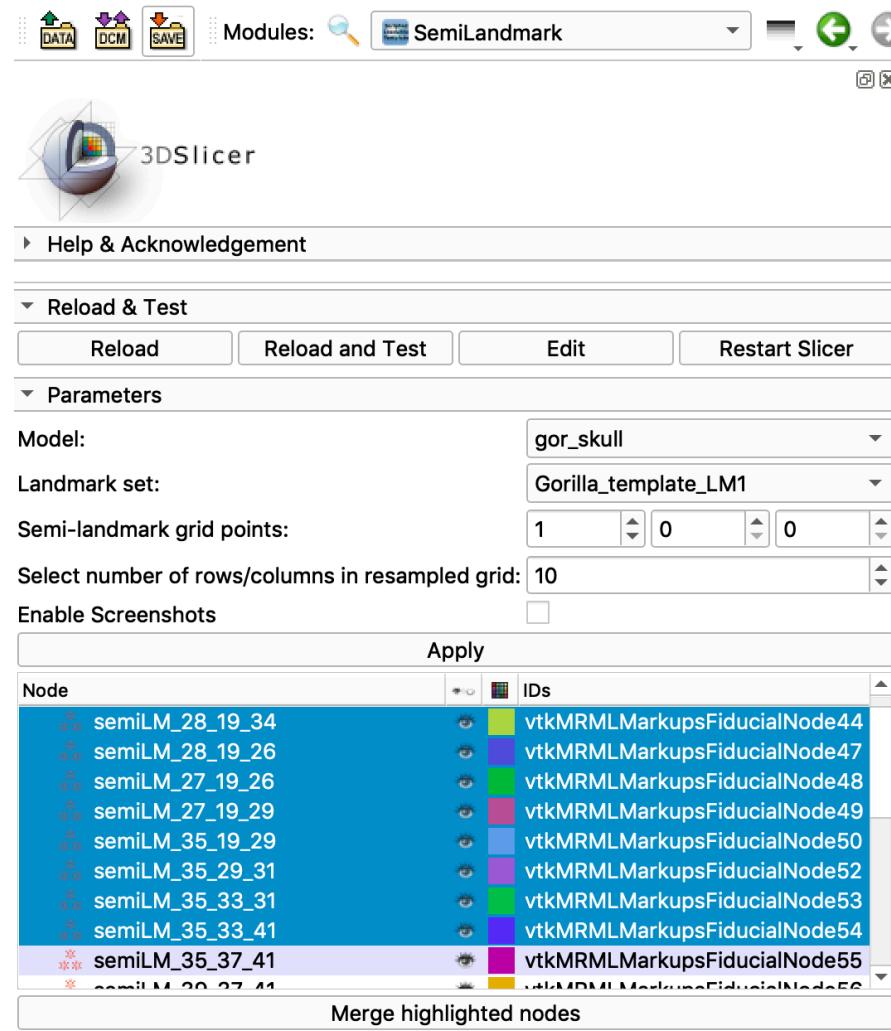
The main workspace displays a 3D model of a skull or mandible. A red grid of points is overlaid on the facial surface of the model. A pink coordinate system is shown at the top of the image. At the bottom, there are sliders for translation along the R (red), Y (yellow), and G (green) axes, with values S: 0.000mm, Y: 0.000mm, and R: 0.000mm. On the left side of the interface, there is a sidebar with sections for Help & Acknowledgement, Reload & Test (with Reload, Reload and Test, Edit, and Restart Slicer buttons), Parameters, Landmark Directory (set to /Users/srolfe), and other settings like Mesh Directory and Output Directory. A yellow diamond-shaped warning sign icon with a black silhouette of a person working on a slope is overlaid on the left side of the interface.

**SLICERMORPH**

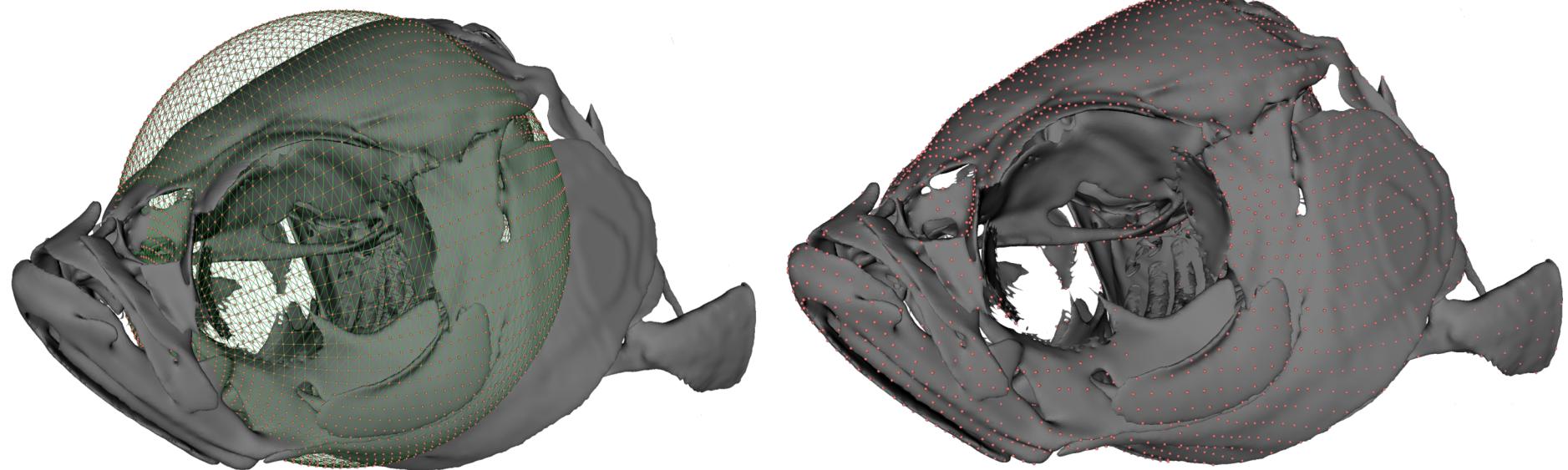
# New Support in SlicerMorph for Semi-landmarking: Curves



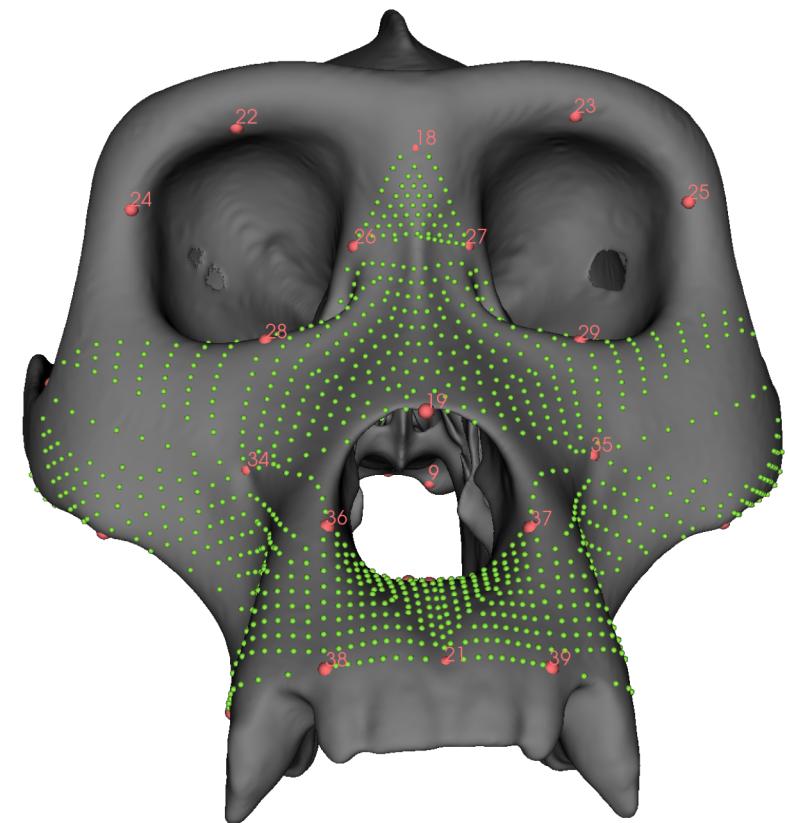
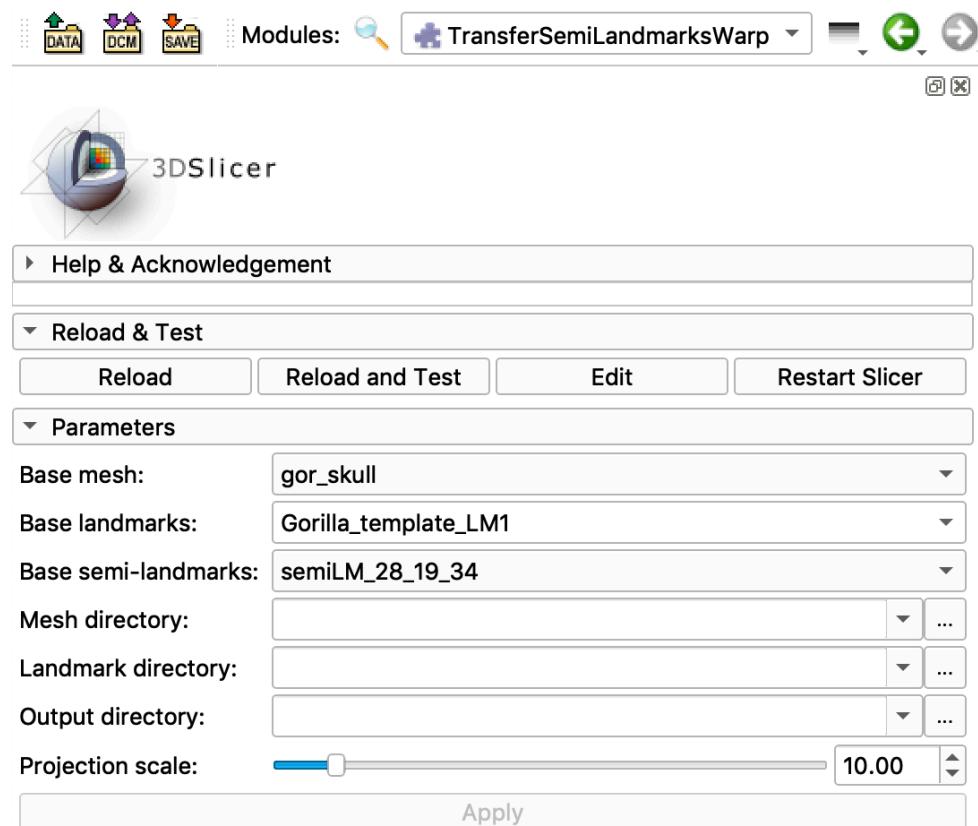
# Patch-based placement of semi-landmarks



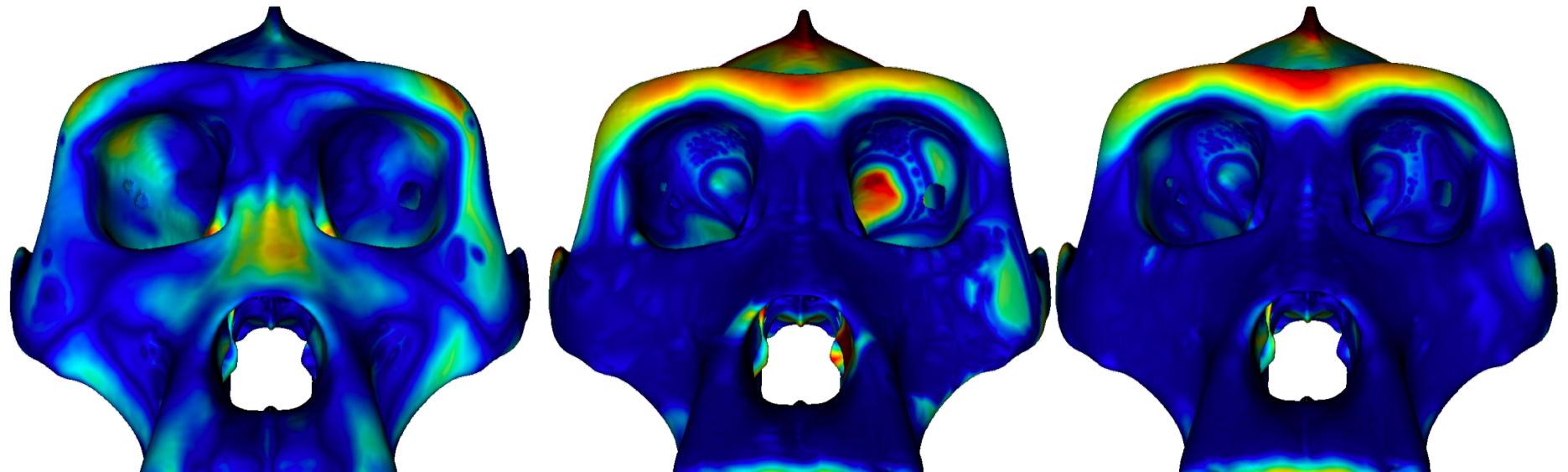
# Coming soon: Spherical sampling



# Transferring Semi-landmarks



# Comparing Semi-landmark methods – warp to mean shape

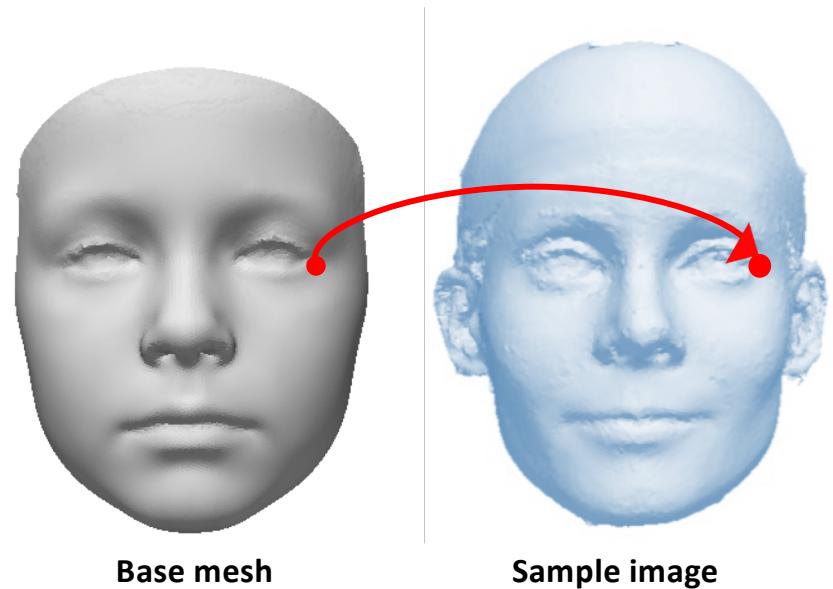


	Original	Patch-placed	Warp-placed
Average			
RMSE	1.27	1.18	1.08



# Deformation-based morphology (DBM)

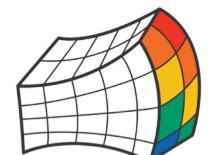
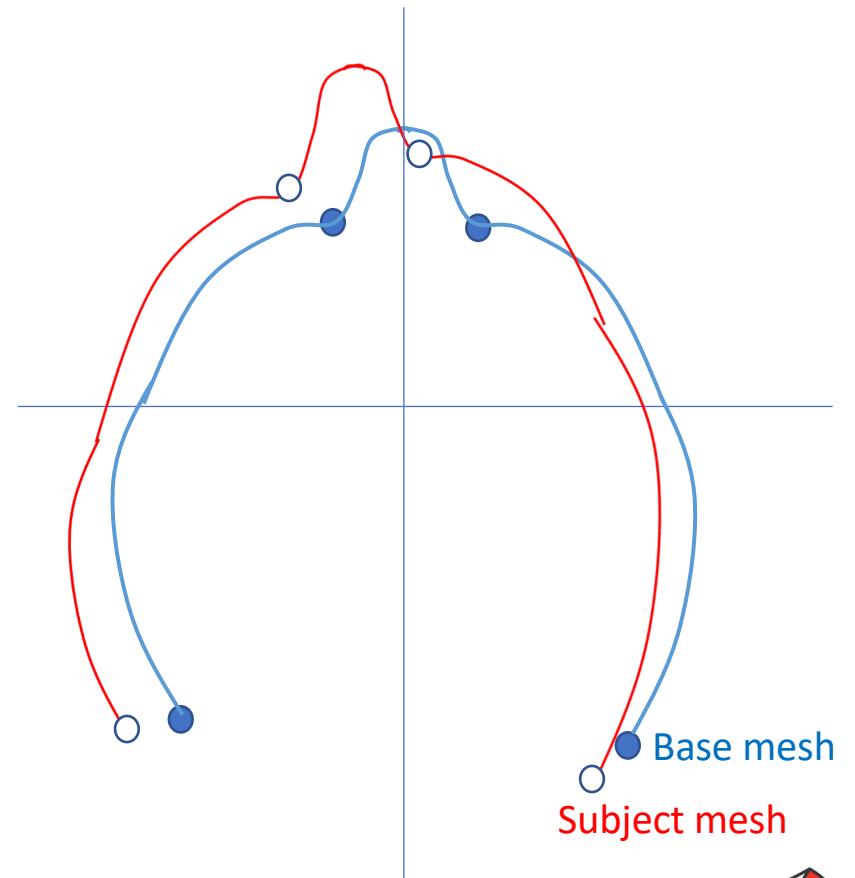
- Calculate or select a base mesh
- Establish point correspondences between the base mesh and all meshes in data set
- Vertices of the base mesh define a new set of dense surface landmarks



*Hutton, T. J., Buxton, B. F., Hammond, P. (2001). "Dense surface point distribution models of the human face," in Proceedings IEEE Workshop on Mathematical Methods in Biomedical Image Analysis (MMBIA 2001) (Kauai, HI).*

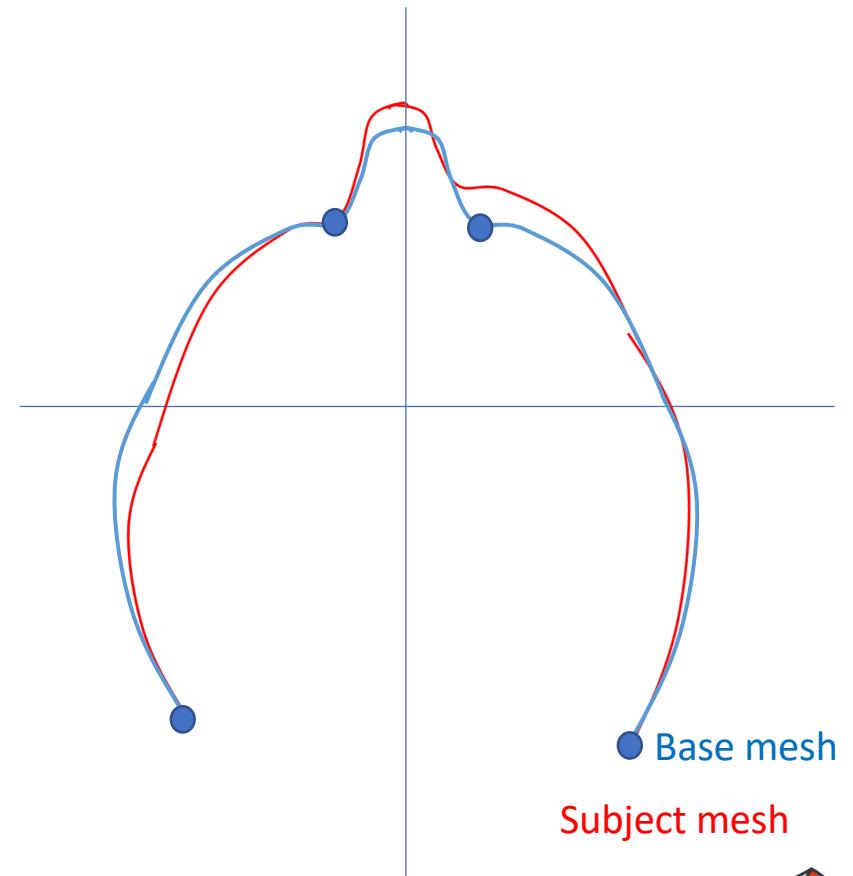


# Steps to find point correspondence



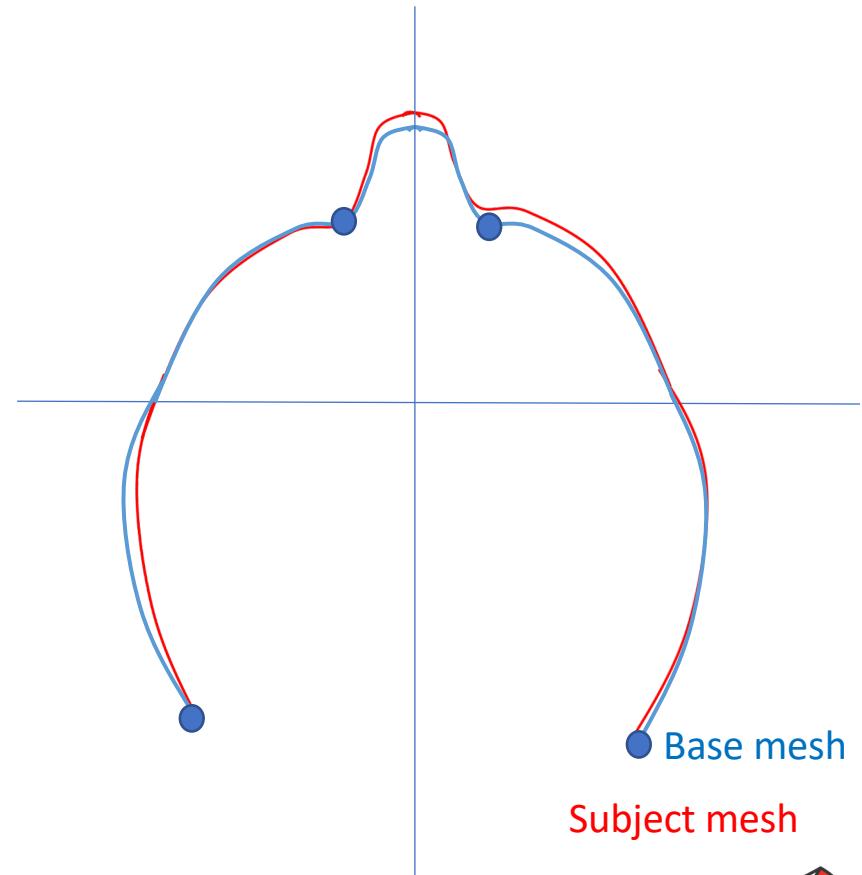
# Steps to find point correspondence

1. Align anatomical landmarks with Generalized Procrustes Analysis (GPA)



# Steps to find point correspondence

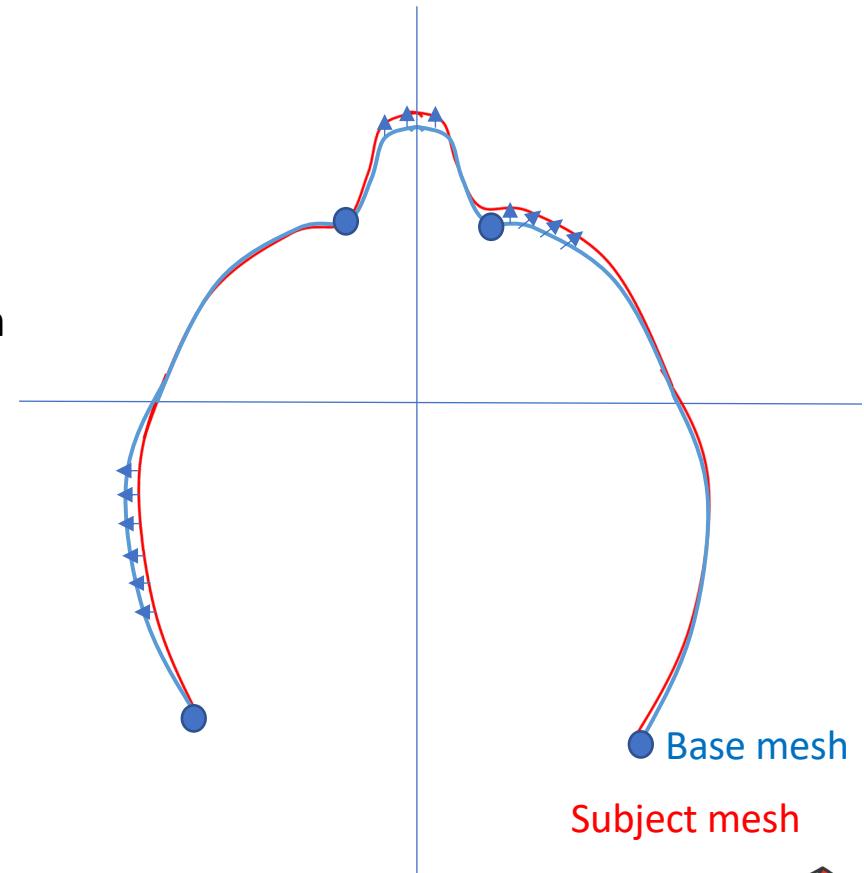
1. Align anatomical landmarks with Generalized Procrustes Analysis (GPA)
2. Warp the meshes to a base mesh with a thin-plate spline (TPS) transformation



SLICERMORPH

# Steps to find point correspondence

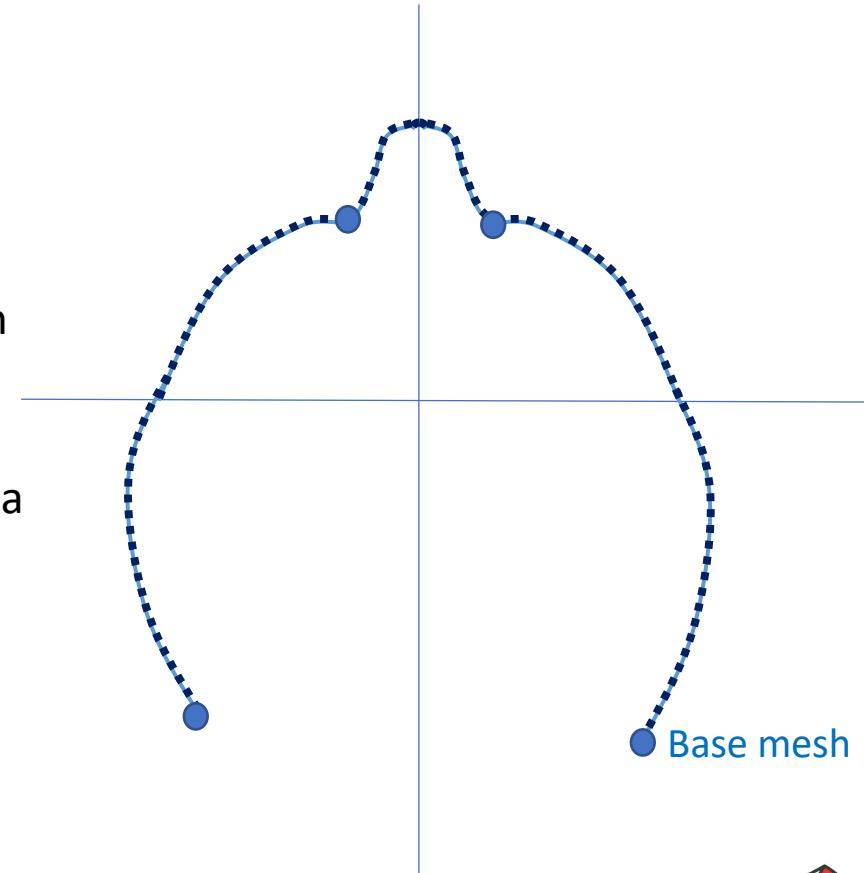
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3. Determine point correspondences for each vertex the iterative closest point (ICCP) algorithm



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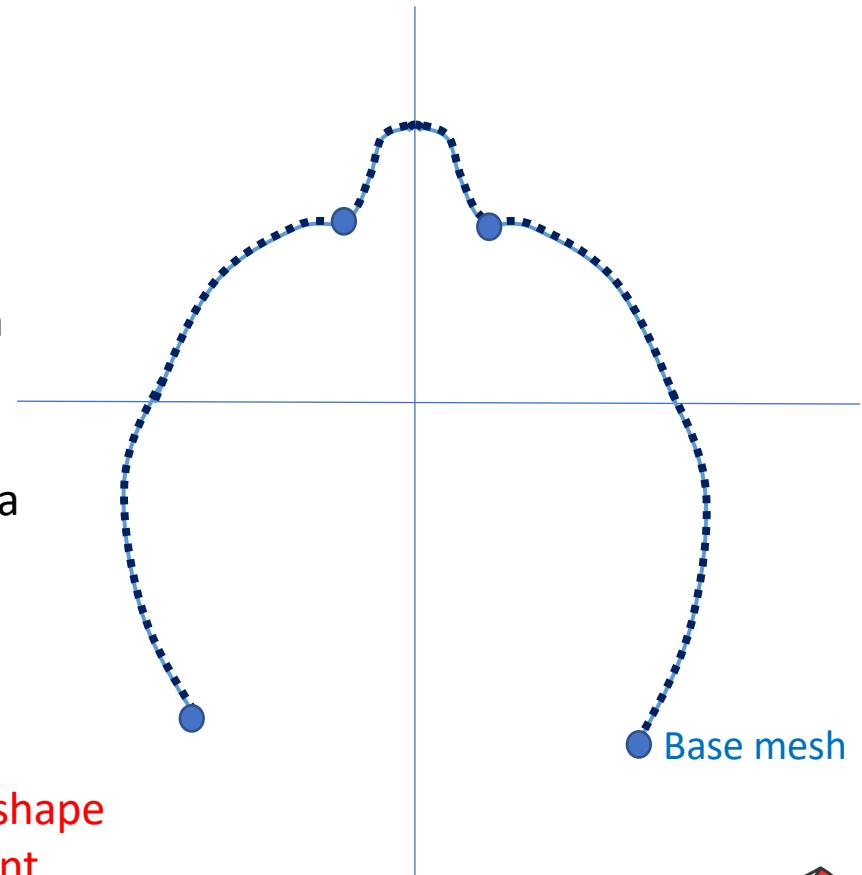
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4. Each vertices of the base mesh represents a landmark point with a known correspondences for each subject



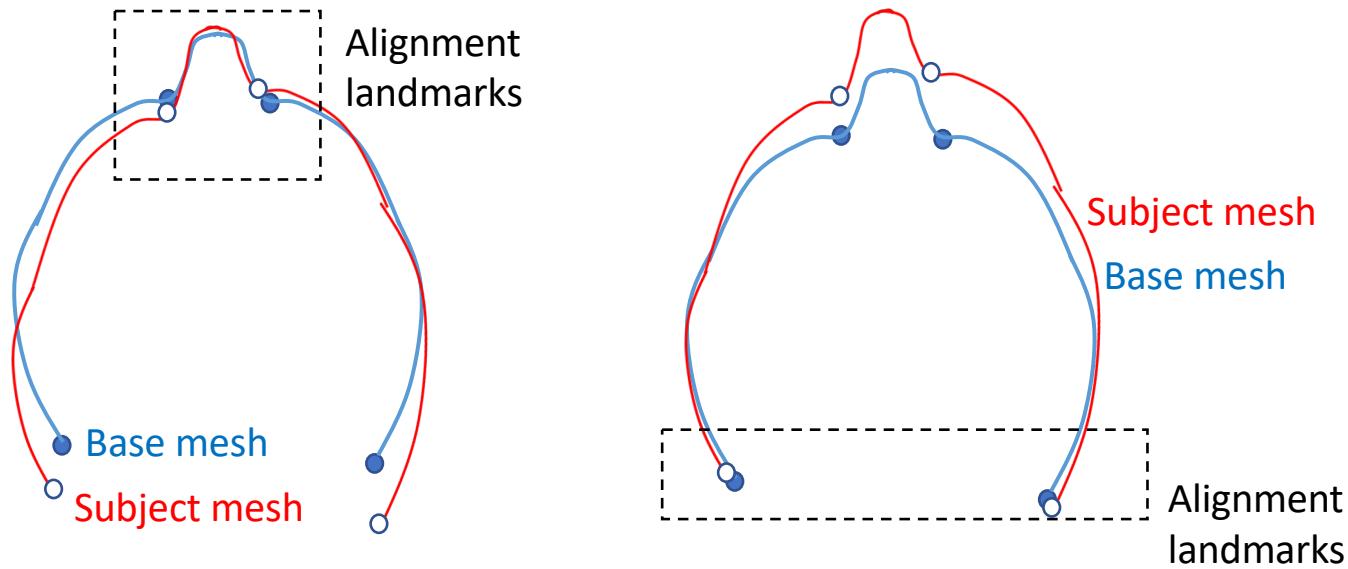
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  - The TPS warped subjects need to be a reasonable approximation of the mean shape
  - This depends on the number of alignment landmarks and the variation in the data set

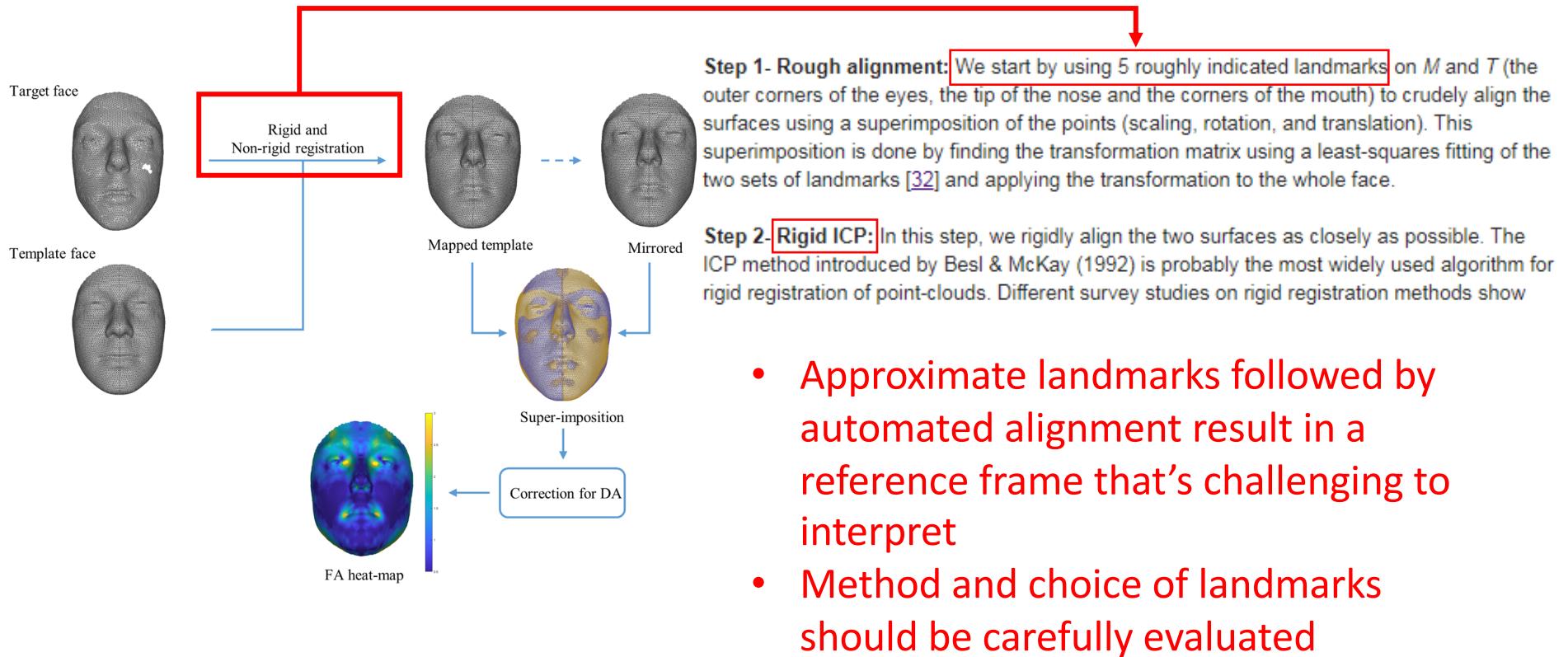


# DBM is not “landmark-free”

- Initial alignment defines the frame of reference
- Choice of landmarks for alignment will impact the interpretation of the results
- Alignment landmarks with known homology will provide an interpretable framework for shape change measured



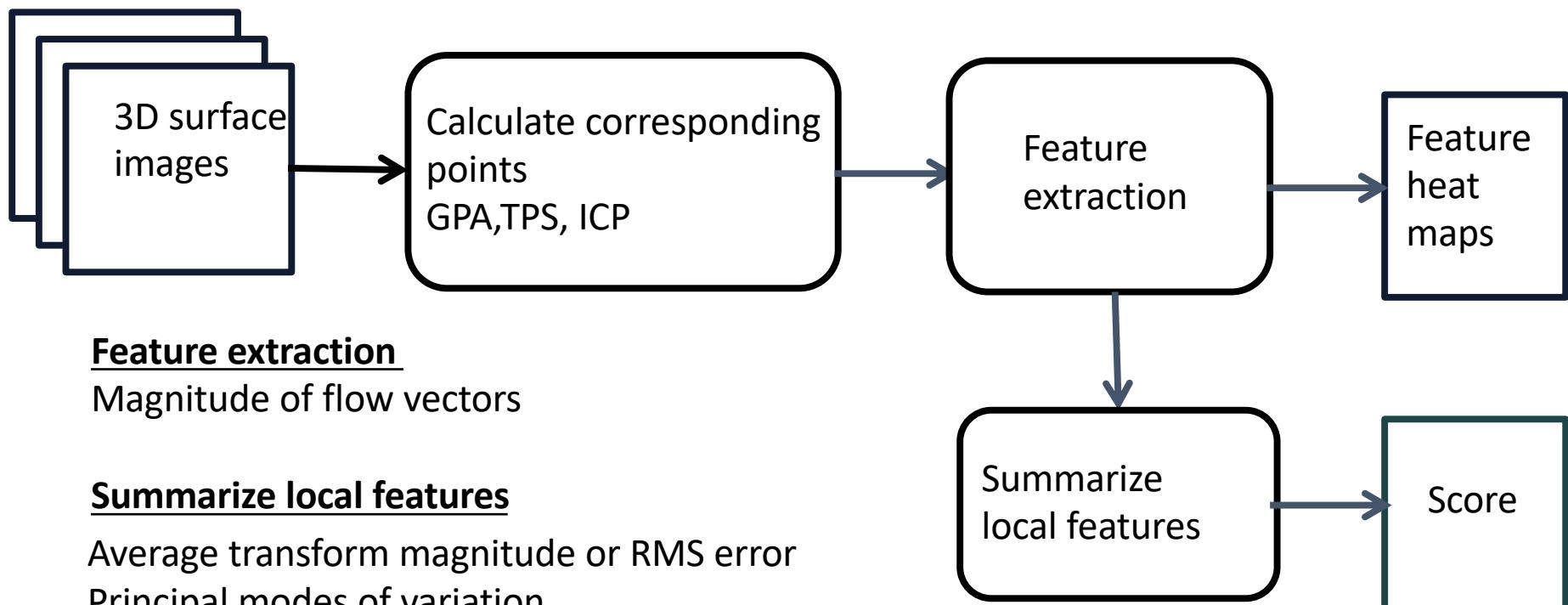
# Initial alignment in DBM



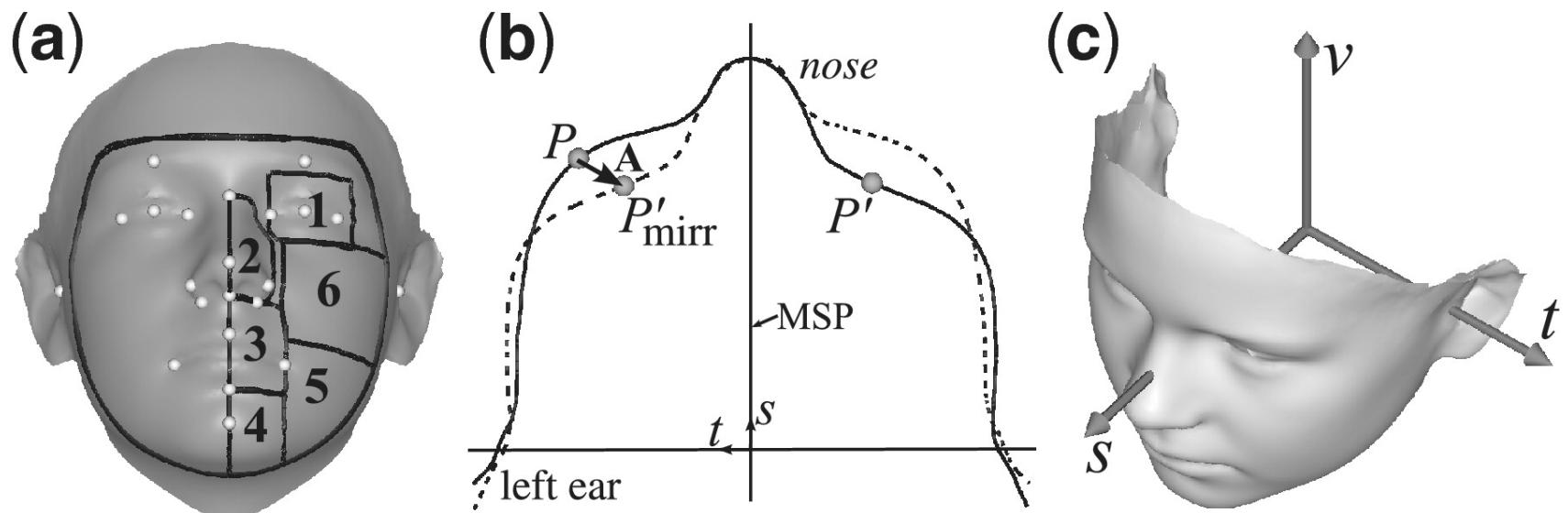
Ekrami, Omid, et al. "Measuring asymmetry from high-density 3D surface scans: An application to human faces." *PloS one* 13.12 (2018): e0207895.



# Quantifying facial shape with Dense Surface Models



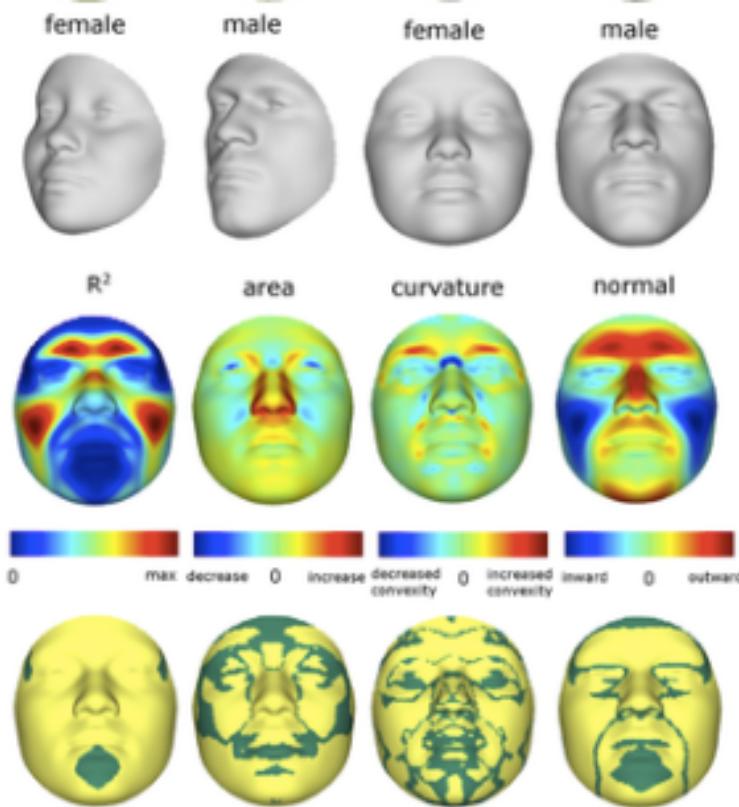
# Automated Quantification of facial asymmetry



Darvann, Tron A., et al. "Automated quantification and analysis of facial asymmetry in children with arthritis in the temporomandibular joint." 2011 *IEEE International Symposium on Biomedical Imaging: From Nano to Macro*. IEEE, 2011.



# BRIM: bootstrapped response-based imputation modeling



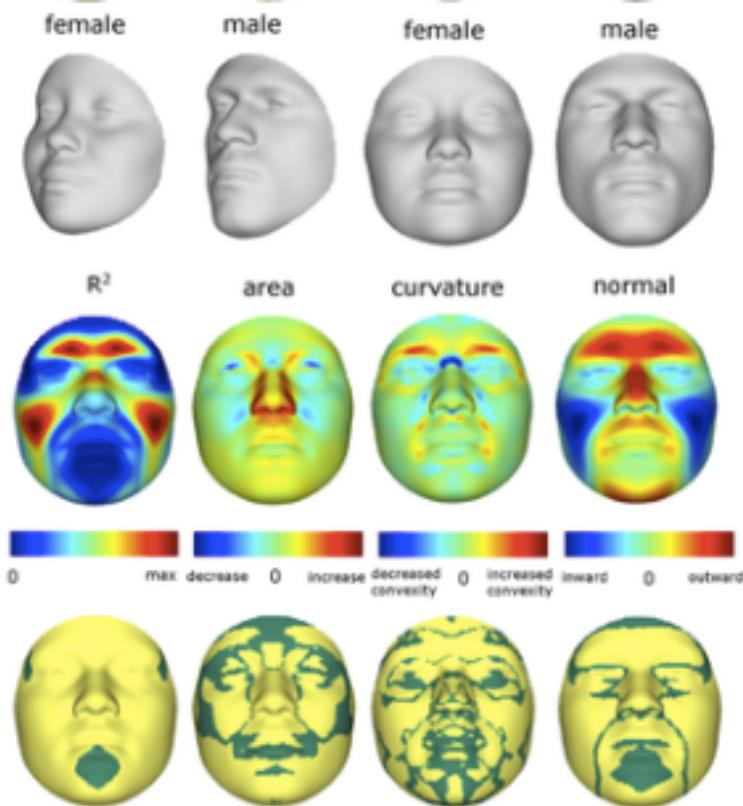
- Technique to investigate the genetic basis for variation in the shape of the human face
- Estimation of a single quantitative axis of variation correspond to effects of single genes, ancestry, or sex

Claes, Peter, et al. "Modeling 3D facial shape from DNA." *PLoS genetics* 10.3 (2014): e1004224.

Hallgrímsson, Benedikt, et al. "Let's face it—complex traits are just not that simple." *PLoS genetics* 10.11 (2014): e1004724.



# BRIM: bootstrapped response-based imputation modeling



- Technique to investigate the genetic basis for variation in the shape of the human face
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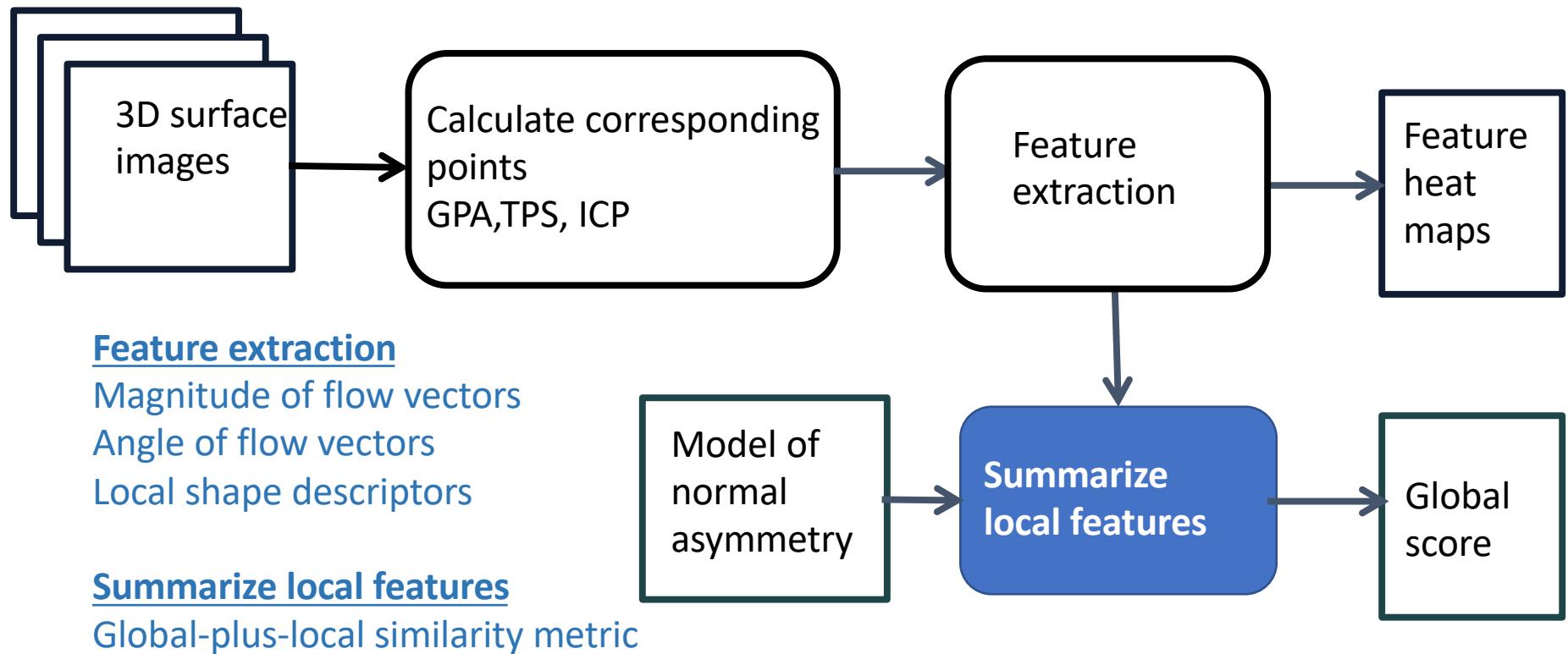
**“...our methods provide the means of identifying the genes that affect facial shape and for modeling the effects of these genes to generate a predicted face.”**

Claes, Peter, et al. "Modeling 3D facial shape from DNA." *PLoS genetics* 10.3 (2014): e1004224.

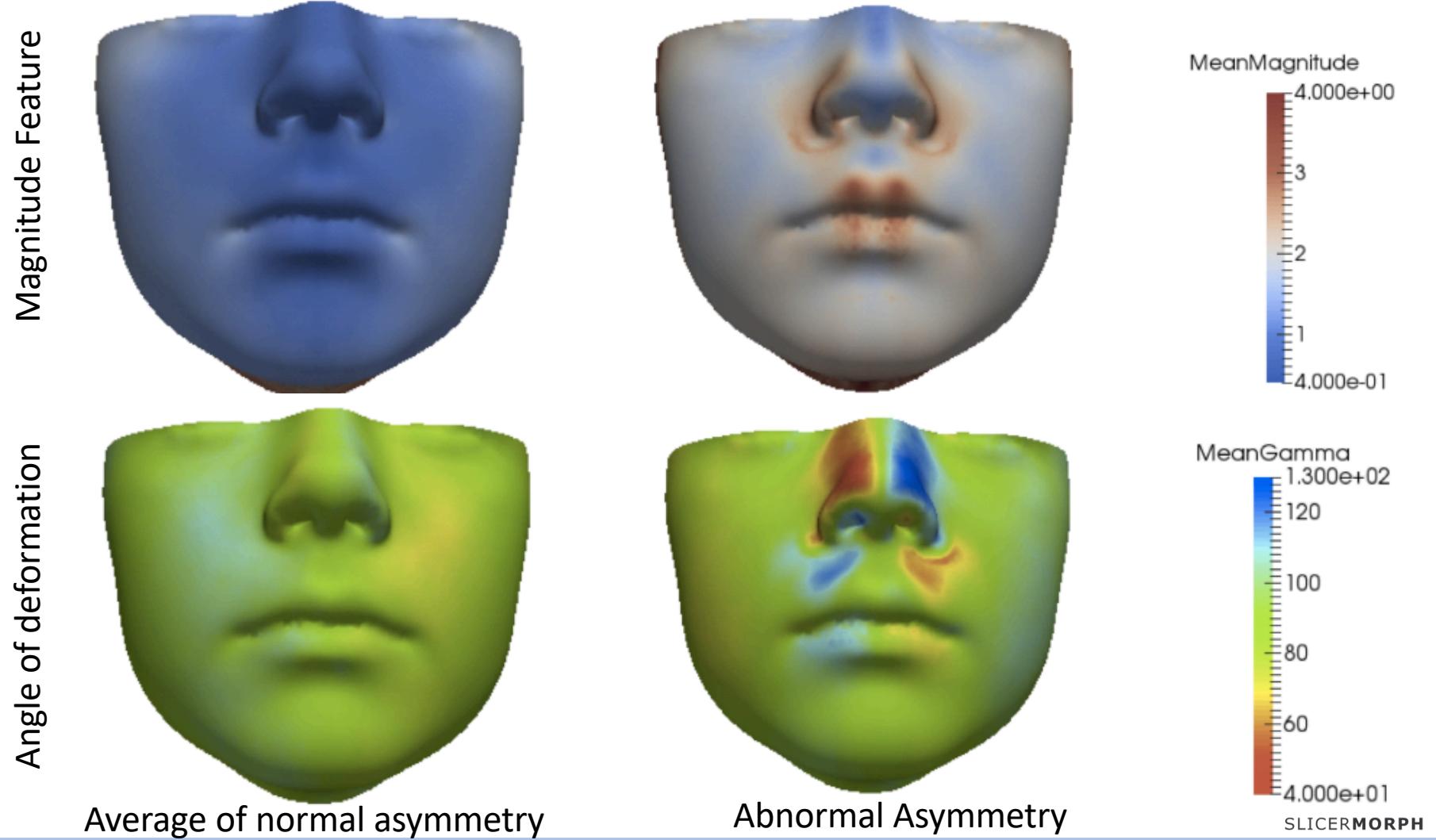
Hallgrímsson, Benedikt, et al. "Let's face it—complex traits are just not that simple." *PLoS genetics* 10.11 (2014): e1004724.



# Summarizing local features

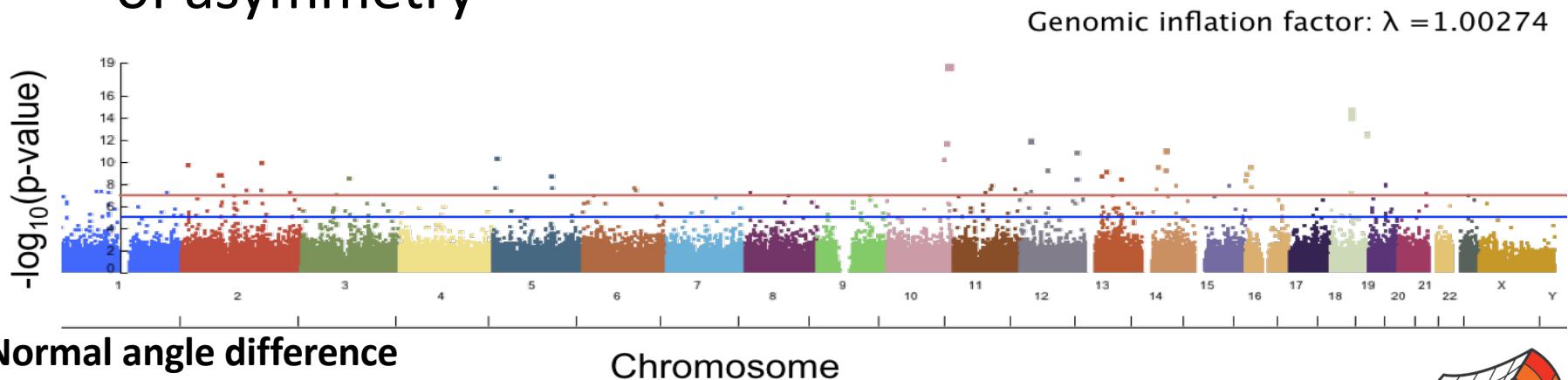


# Quantification of abnormal asymmetry



# Genetic basis of facial asymmetry

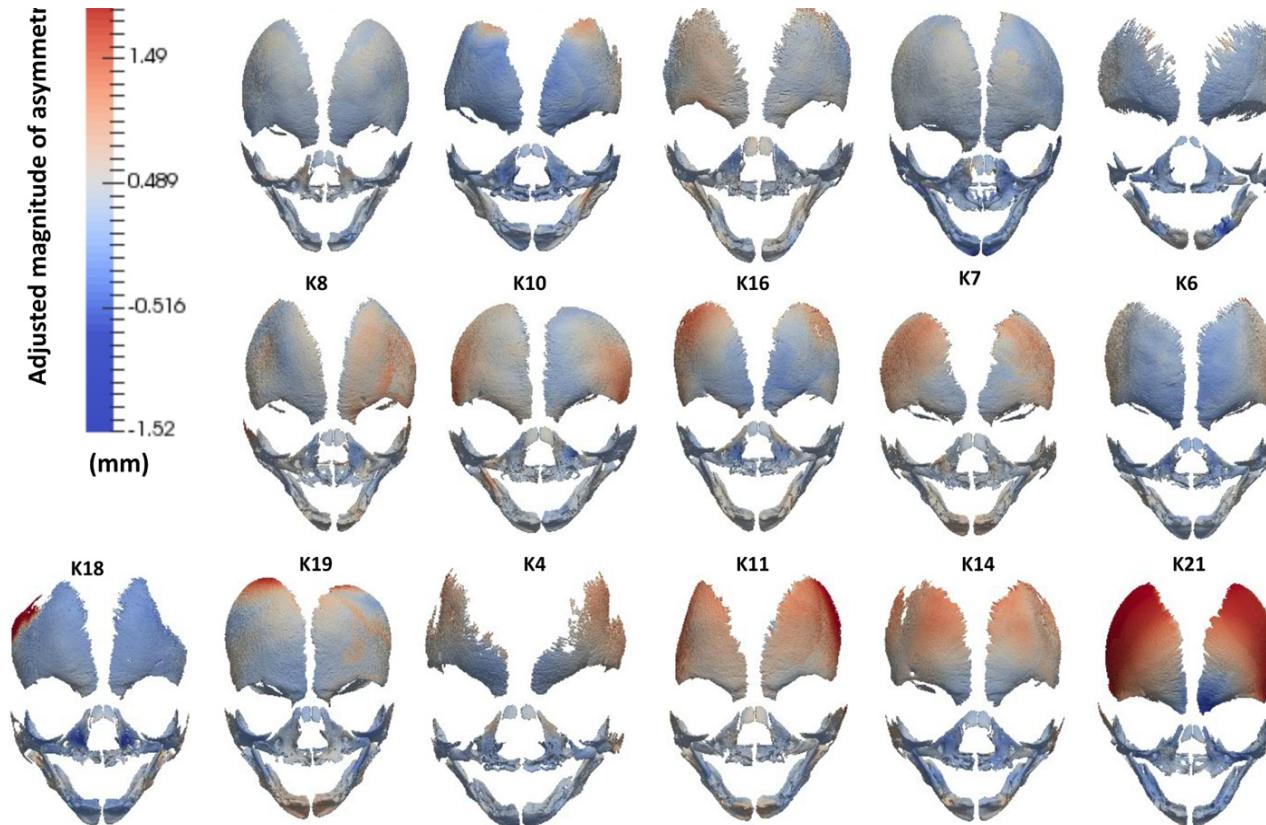
- Aim to use automated phenotyping to produce a score of facial asymmetry that can summarize local deviations from normal asymmetry
- Correlate with SNP data to investigate genetic basis of asymmetry



Rolfe, Sara, Su-In Lee, and Linda Shapiro. "Associations between genetic data and quantitative assessment of normal facial asymmetry." *Frontiers in genetics* 9 (2018): 659.



# Analysis of asymmetry in developing facial bones



Katsube, M., Rolfe, S.M., Bortolussi, S.R., Yamaguchi, Y., Richman, J.M., Yamada, S. and Vora, S.R., 2019. Analysis of facial skeletal asymmetry during foetal development using  $\mu$ CT imaging. *Orthodontics & craniofacial research*, 22, pp.199-206.

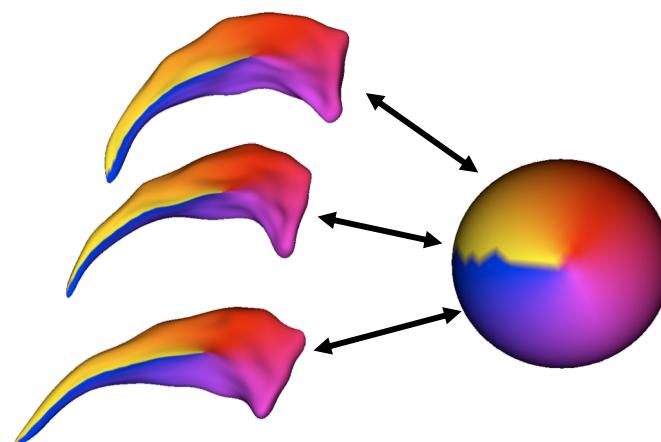
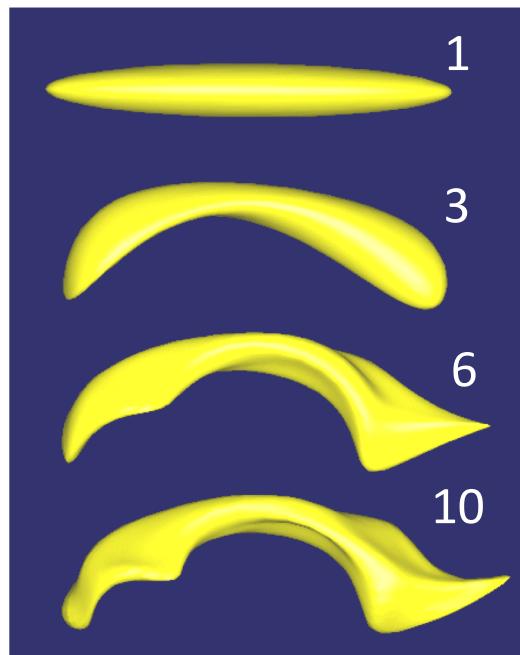


# Limitations of DBM

- No guarantee of homology
- Results need to be interpreted with respect to initial alignment
- High dimensional data output
- Computationally intensive for large datasets



# Spherical harmonic representation and point distributed models (SPHARM-PDM)

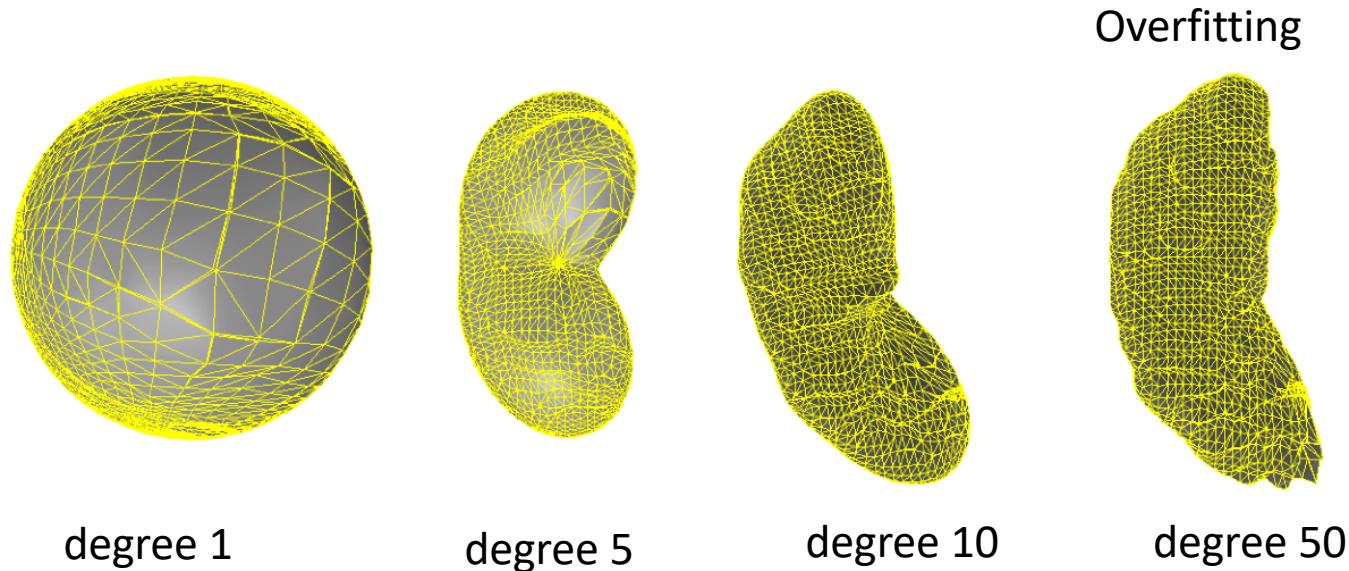


SPHARM  
description is  
computed from  
the mesh and its  
spherical  
parameterization

Styner, Lieberman, Pantazis, Gerig: Boundary and Medial Shape Analysis of the Hippocampus in Schizophrenia, Medical Image Analysis, 2004, pp 197-203



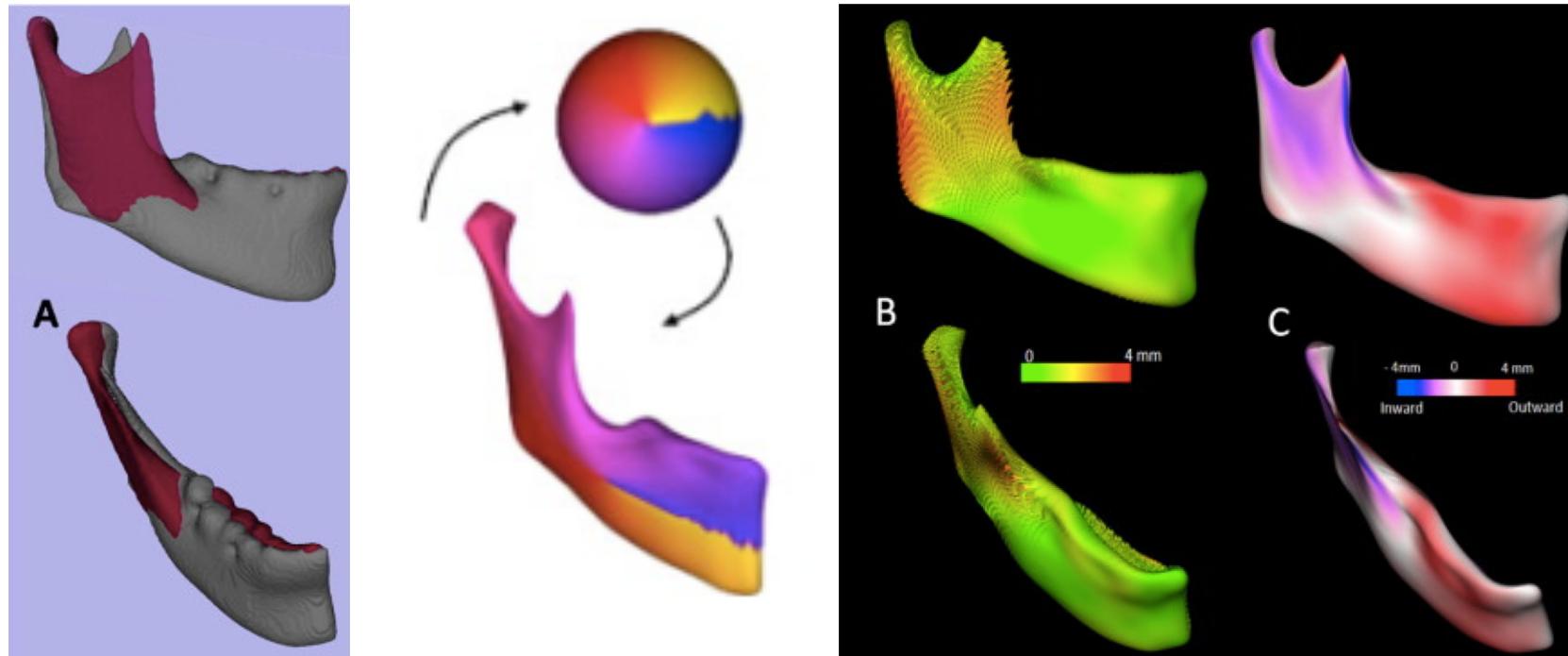
# SPHARM modeling



- The SPHARM description is then sampled into a triangulated surface
- Triangulated surfaces with correspondences are computed via icosahedron of the spherical parameterization



# SPHARM analysis of mandibular asymmetry



Cevidanes, Lucia HS, et al. "Three-dimensional quantification of mandibular asymmetry through cone-beam computerized tomography." *Oral Surgery, Oral Medicine, Oral Pathology, Oral Radiology, and Endodontics* 111.6 (2011): 757-770.

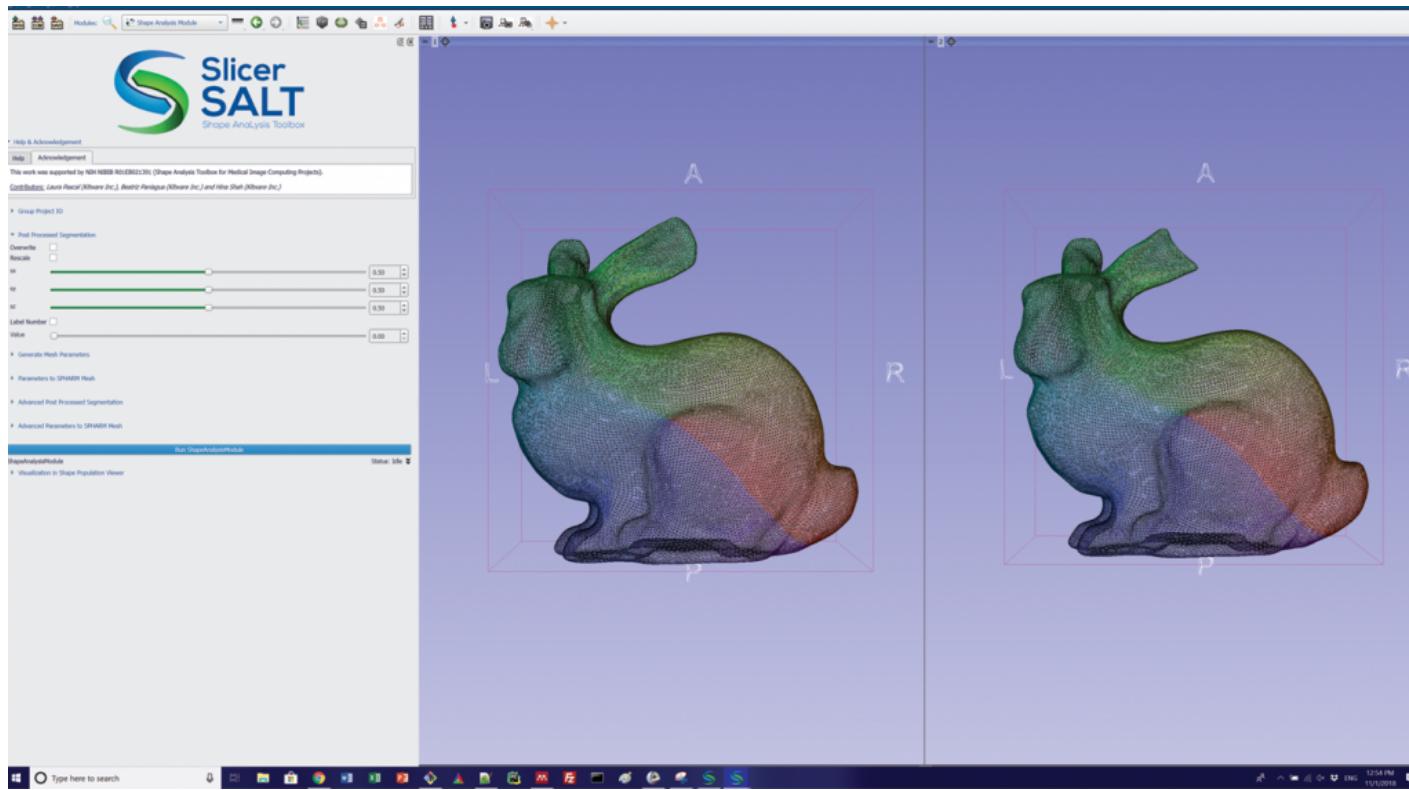


# Limitations of SPHARM methods

- Requirement for spherical topology of meshes
- Challenging for complicated or noisy data sets
- Developed for high N of shapes with similar morphology – should perform a quality assessment to check for correspondence issues



# Slicer SALT: Shape AnaLysis Toolbox

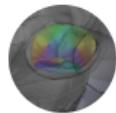


Compute point distributed models (pdm) using spherical harmonic representation (SPHARM-PDM)

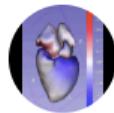


# Coming soon

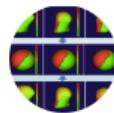
With SlicerSALT you can...



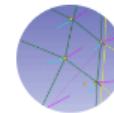
Compute Point Distributed Models (PDM) using Spherical Harmonic Representation (SPHARM-PDM)



Run 4D regression in a collection of 3D PDMs associated to a linear variable (i.e. age)



Perform correspondence optimization using study-wise shape analysis



Fit skeletal representations (s-reps) to a collection of binary volumes



Compute image-based correspondence in binary volumes of non-spherical and complex topologies



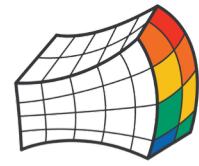
Use advanced shape statistics for scientific hypothesis testing



Access all functionality through the graphical user interface



Perform command line batch processing



# Questions?



# Resources

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