## ShapeWorks

The tutorial

## Correspondence Pipeline

## Align $\rightarrow$ Antialias $\rightarrow \underset{\text { Transform }}{\text { Distance }} \rightarrow$ Initialize $\rightarrow$ Optimize $\rightarrow$ Align $\boldsymbol{I}$ <br> Align $\rightarrow$ Antialias $\rightarrow \underset{\text { Transform }}{\text { Distance }} \rightarrow$ Initialize $\rightarrow$ Optimize $\rightarrow$ Align $\boldsymbol{I}$



## Command Line Tools



Analysis: PCA, Group differences ShapeWorksView

## ShapeWorksGroom

Command line collection of preprocessing filters

## Syntax

ShapeWorksGroom torus.preprocess1.params isolate hole_fill center auto_crop
ShapeWorksGroom torus.preprocess2.params antialias fastmarching blur

Basic filters available

- isolate
- hole fill
- antialias
- fastmarching
- blur

DEMO: torus example

## ShapeWorksGroom



Example tori shapes from population parameterized by $r$ and $R$

## ShapeWorksRun

Command line tool to initialize and optimize particle positions on shapes in the ensemble

## Syntax

ShapeWorksRun torus.correspondence. params

Notable parameters

- \# particles
- adaptivity
- alignment


## DEMO: torus example

## ShapeWorksRun



Correspondences overlaid on 3 of the input shapes

## ShapeWorksView

GUI to visualize correspondences and perform statistical analysis

Syntax
ShapeWorksView torus.analyze.params
Notable parameters

- Reconstructed shapes
- Modes of variation
- Group differences (not in demo, only if 2 populations are available)


## DEMO: torus example

## ShapeWorksView


mean


Modes of variation captured by the correspondence model.
The first (top) and second (bottom) modes capture the shape variation consistent with the generative model

## Recent Workt

## Challenges of Nonregular Shapes


(a)
(b)

Fig: Incorrect correspondences near sharp features when (a) points with different tangent spaces interact, (b) nearby points sampling different parts of the surface interact, (c) optimization is based only on point positions on different shapes
$\square$ Geodesic distances
[ computed using intermediate triangular meshes
[ pre-computed between vertices [Fu, et. alt ${ }^{\dagger}$ ]
otherwise computed using two-layered Barycentric interpolation
$\square$ Surface normal entropy
$\square$ penalize divergence from "mean" in the space of surface normals
$\square$ helps disambiguate correspondences near convoluted features

[^0]
## Group Comparison: LV wall $\dagger$

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Mean differences between normal and ischemic groups (blue => expansion, yellow => contraction)
Top: PBM ${ }^{*}$, Bottom: proposed method
$\square$ Study shape differences between normal and ischemic LV wall segmented at end diastole (ED)
$\square$ Results
[ Group mean differences significant with $p$-value $<0.01$
$\square$ Shape changes spatially consistent with previously published results

## Nonlinear Growth Model



Varying asymptote*



Varying delay*



Varying speed*


Estimate Gompertz* model parameters

Fig: Replacing linear regression model with nonlinear Gompertz model for optimization


Fig: Progression of growth at three time-points from neonate -4 years

## Mixed Effects Model



## Mixed Effects Model: Trends



Group trend


Individual trend

## Pairwise Distance Features

## Pairwise interparticle distance (Euclidean/Geodesic) as a feature for correspondence optimization



Fig: Initial results from study of cortext shapes. Two examples from the population with correspondences overlaid. Note that correspondences are not good near the top of the cortex, suggesting the need for additional features (e.g. curvature) to be included

## Thank you!

Questions?


[^0]:    † Fu, Z., Kirby, M., Whitaker, R.: A fast iterative method for solving the eikonal equation on triangulated meshes. SIAM Journal on Scientic Computing (2011) To appear

