SHREC'16: MATCHING OF DEFORMABLE SHAPES WITH TOPOLOGICAL NOISE

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8th May 2016

3DOR, Lisbon

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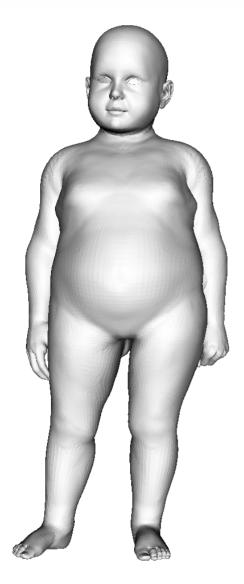
REAL SCANS

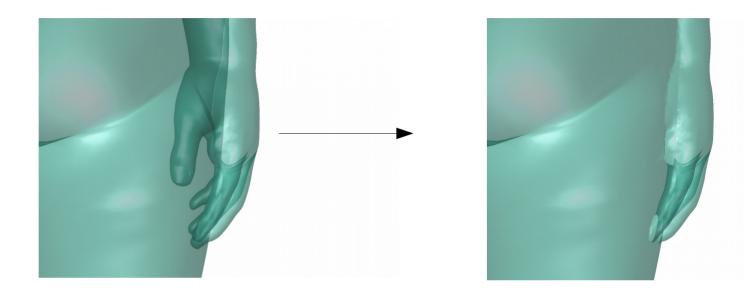




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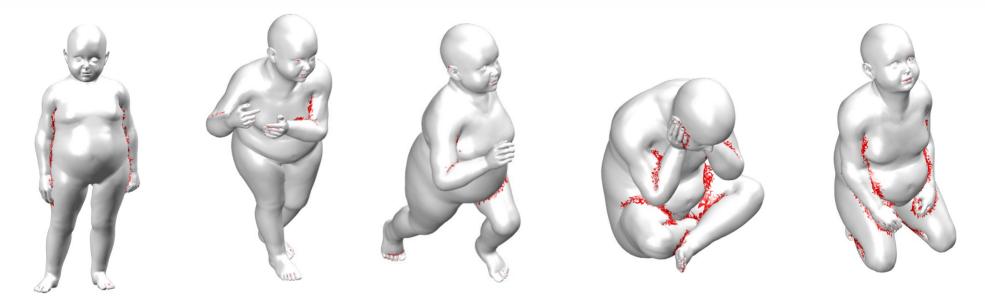
THE DATASET





- Shapes from the KIDS data set
- Merged self-intersecting parts to form the outer hull

THE DATASET



- Two resolutions: ~60-80k and ~10k vertices
- Indicator Map of unmatched points
- Left-Right Map
- Download at in.tum.de/~laehner/shrec2016/

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THE CONTEST

- Task: Provide pointwise correspondences for 90 pairs of shapes of the same class with topological noise
- 15 shapes with ground-truth correspondence to a null shape are given as training data
- Separate evaluation for the high and low resolution data set



EVALUATION

• Following the Princeton protocol

$$\varepsilon(x) = \frac{d_{\mathcal{N}}(y, y^*)}{\operatorname{area}(\mathcal{N})^{0.5}}$$

• How large is the deviation from the ground truth w.r.t. to the diameter of the shape?

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METHODS AND RESULTS

MULTIDIMENSIONAL SCALING (EM)

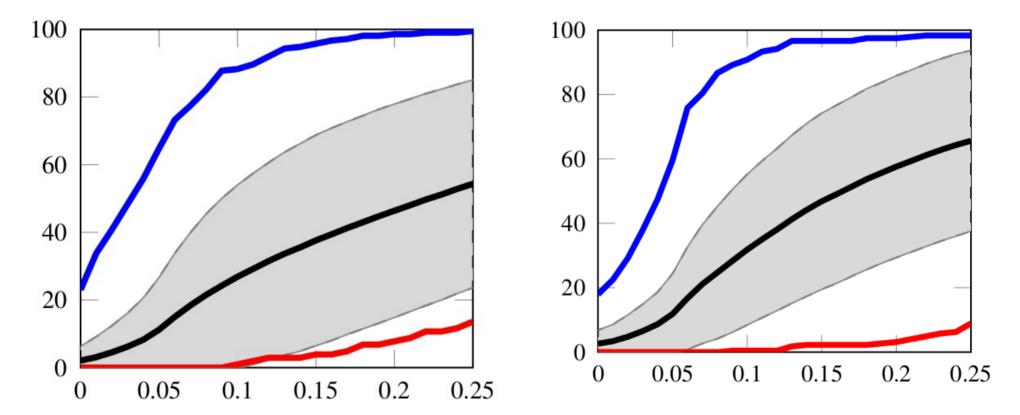
- Based on:
- Sahillioglu Y., Yemez Y.: *Minimum-distortion isometric shape correspondence using EM algorithm*, IEEE Trans. on Pattern Analysis and Machine Intelligence, 2012
- Lipman Y., Rustamov R., Funkhouser T.: *Biharmonic distance* , ACM Transactions on Graphics, 2010



EM: Princeton Protocol

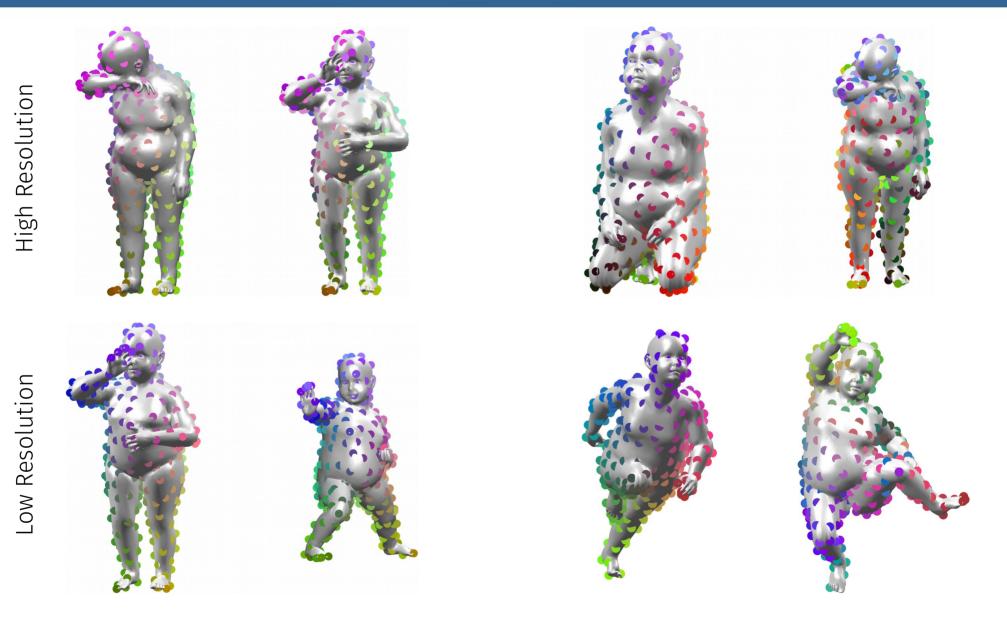
EM (high res)

EM (low res)





EM: Examples



Best Matching

Worst Matching

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GREEN'S FUNCTION (GE)

- Starts with four user-given matches
- Solves Green's function:

$$\Delta g_x = \delta_x$$

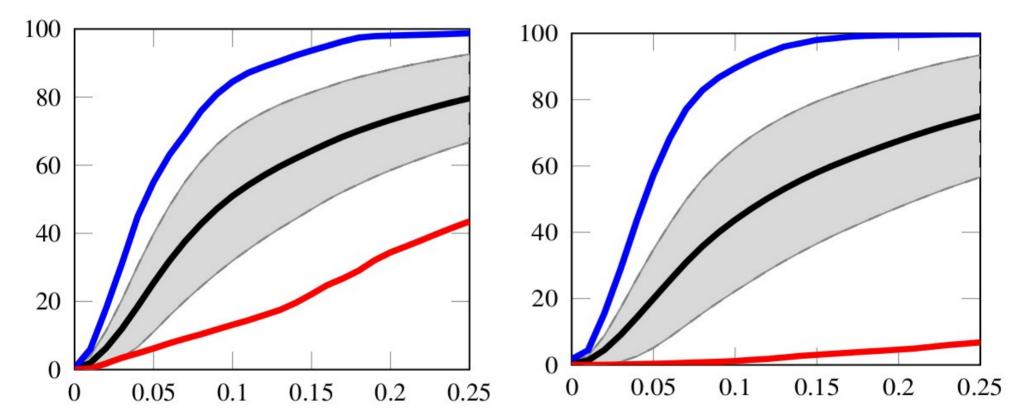
• Participants: Oliver Burghard, Alexander Dieckmann, Reinhard Klein (University of Bonn)



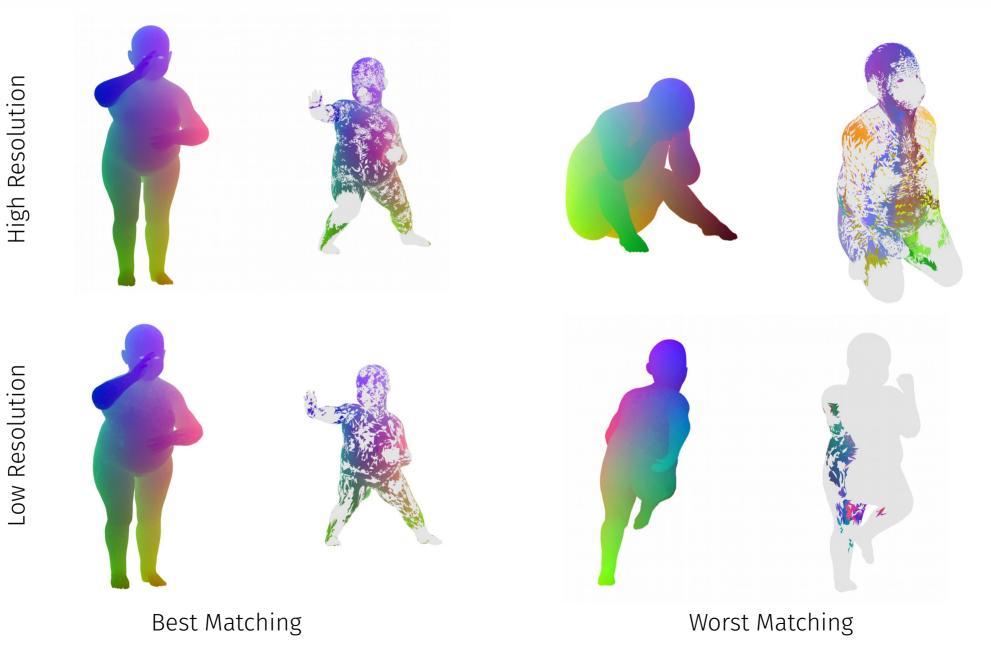
GE: Princeton Protocol

GE (high res)

GE (low res)



GE: Examples



RANDOM FORESTS (RF)

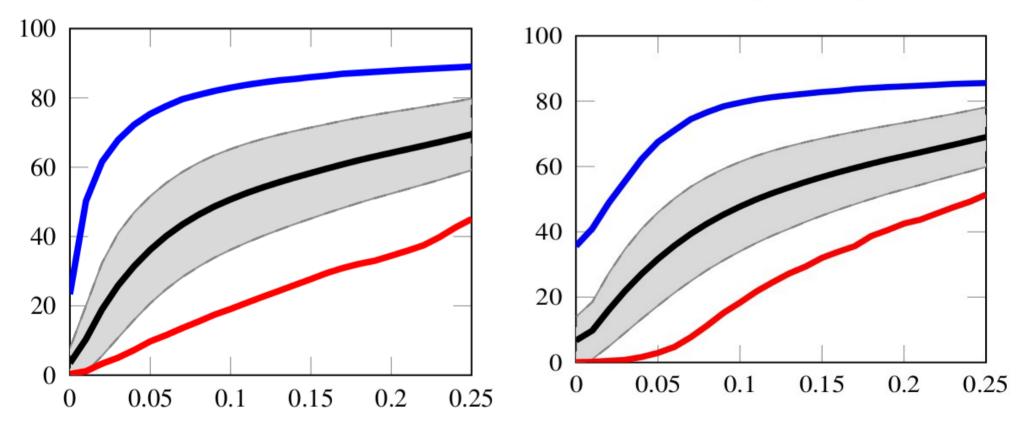
- Uses the training set to train a Random Forest
- Based on: Rodolà E., Rota Bulò S., Windheuser T., Vestner M., Cremers D.: Dense non-rigid shape correspondence using random forests, In Proc. CVPR, 2014
- Sun J., Ovsjanikov M., Guibas L.: A concise and provably informative multi-scale signature based on heat diffusion, In Proc. SGP, 2009



RF: Princeton Protocol

RF (high res)

RF (low res)



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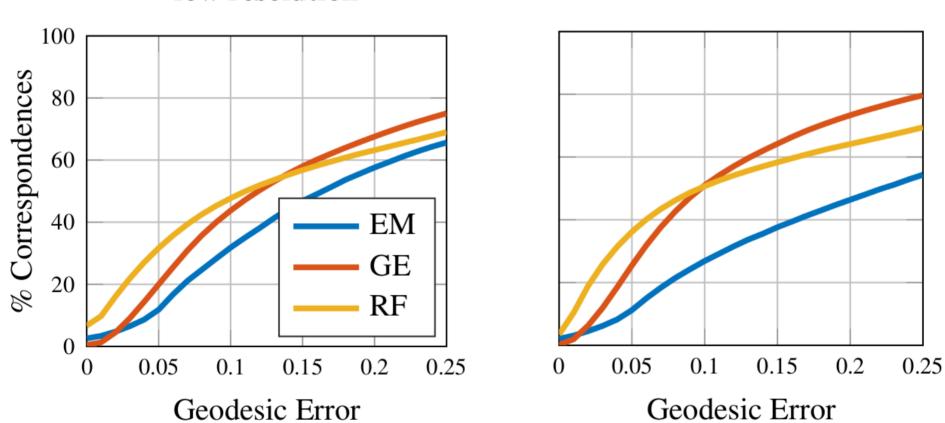
RF: Examples



Best Matching

Worst Matching





high resolution

low resolution



- No method performs really well on average and there is quite some space for improvement
- There are some shapes on which all methods perform badly



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Thank you for your attention!