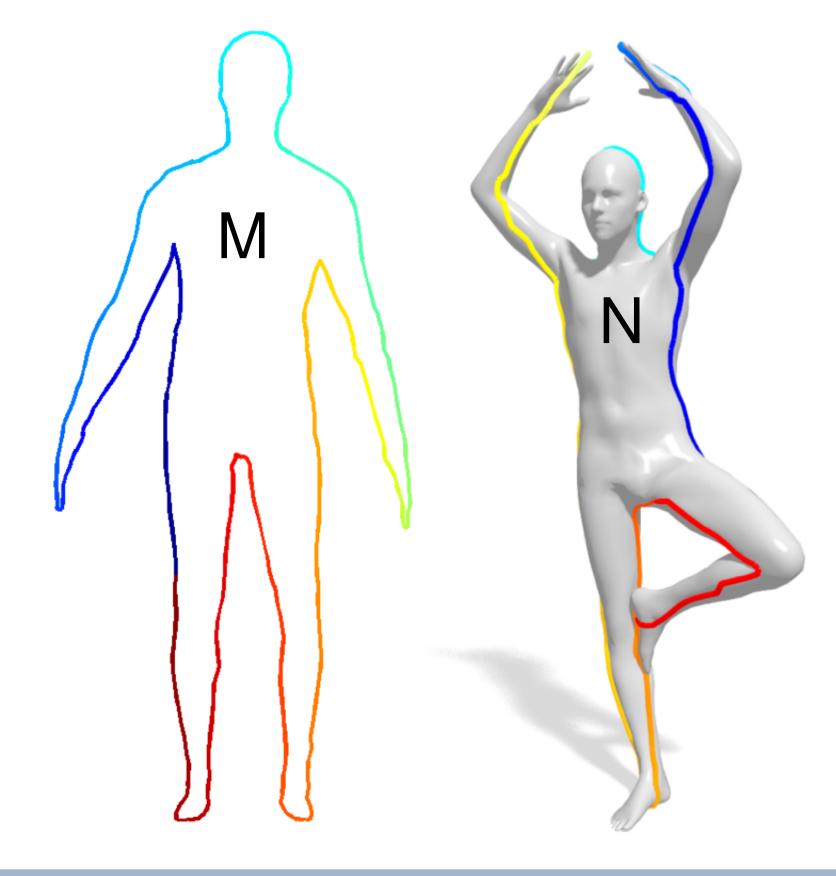


Efficient Globally Optimal 2D-to-3D Deformable Shape Matching

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The Task

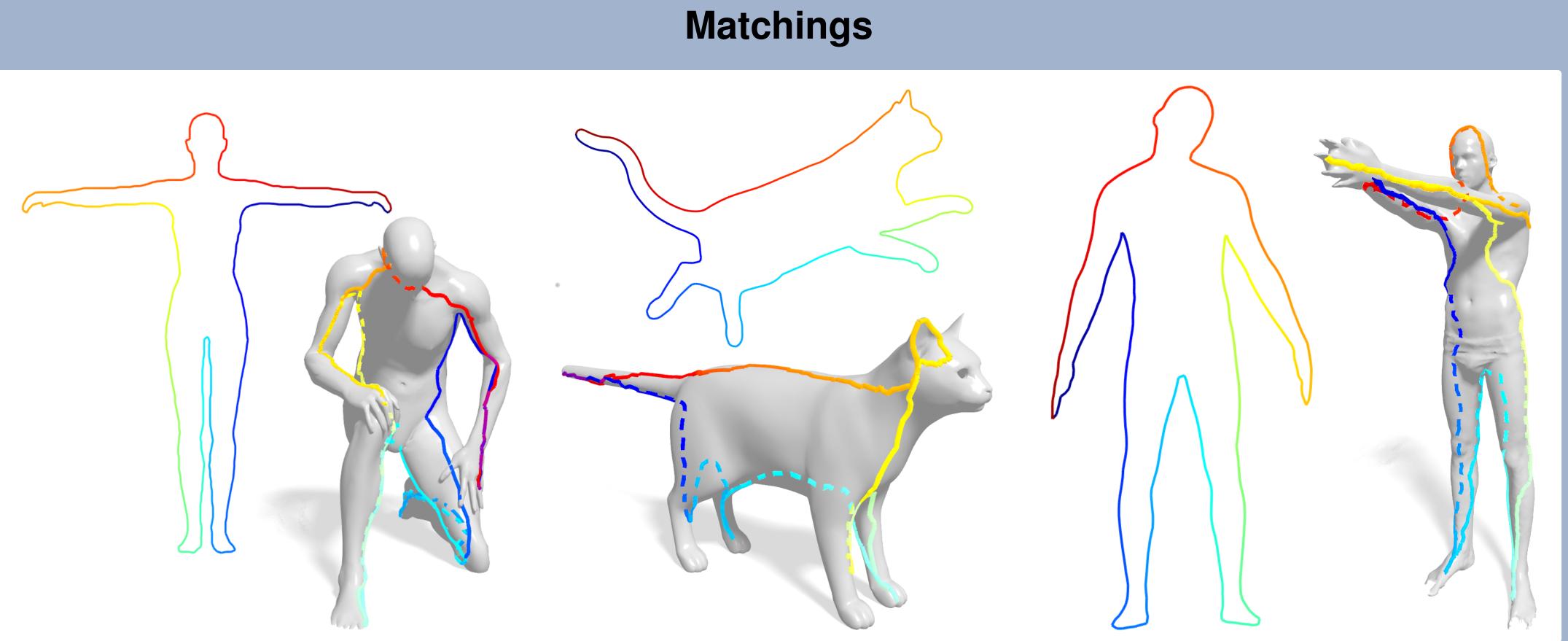
Finding a **continuous** matching $\varphi \colon M \to N$ between a 2D *template shape* M (with m nodes) and a 3D model shape N (with n nodes).



Each node in the product manifold represents a match between a point on M and N. The graph is constructed such that each path through it results in a continuous matching.

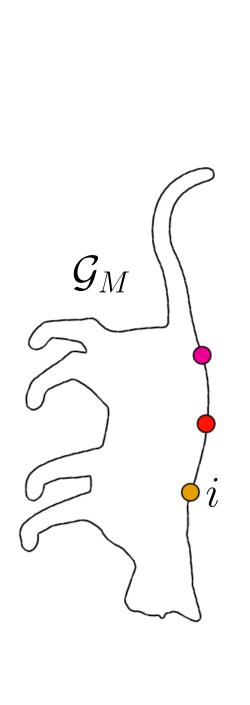
Contributions

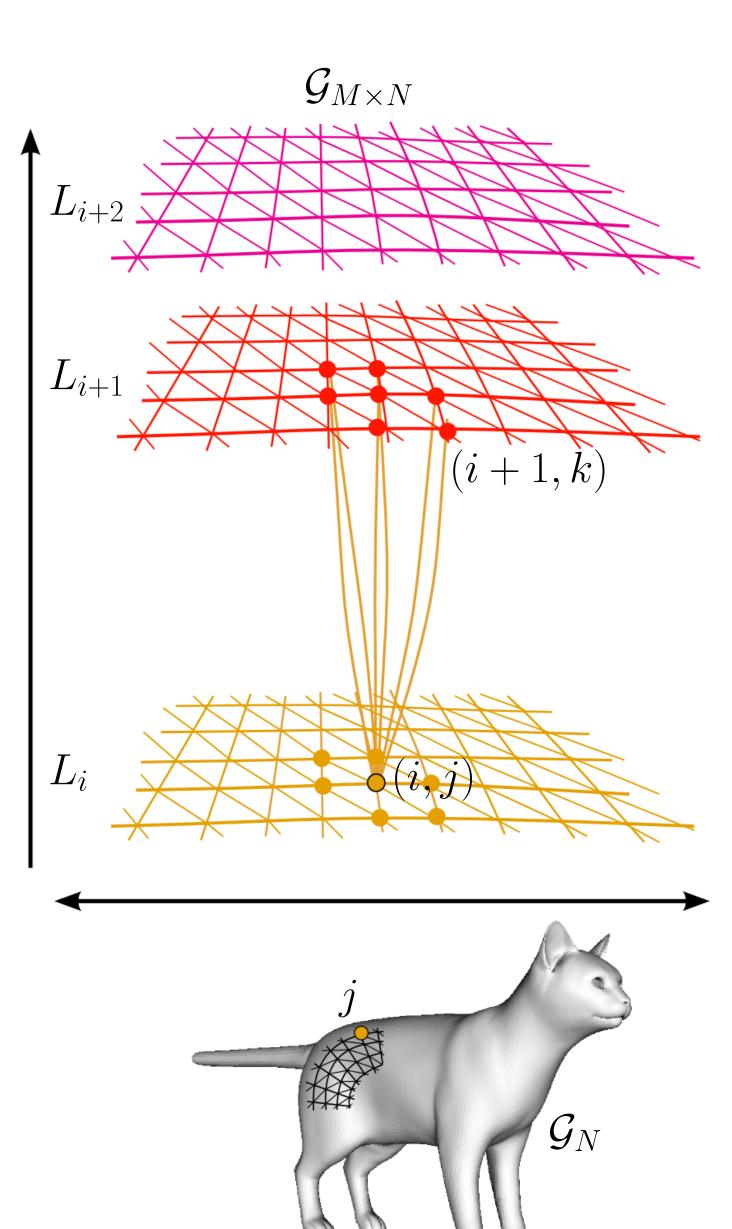
- Fully automated Shape Matching method
- Continuous matching
- Globally optimal solution
- Runtime: $\mathcal{O}(mn^2 \log(n))$
- Comparable 2D and 3D features



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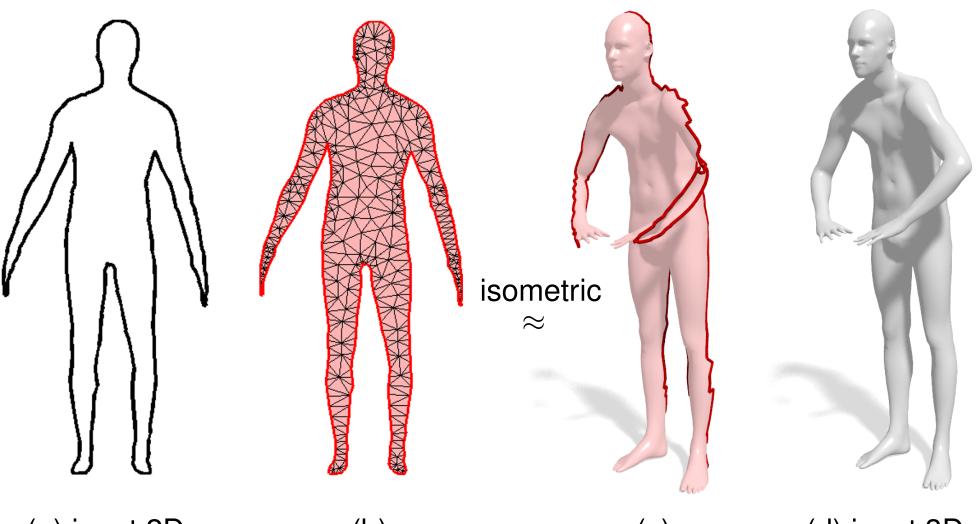
Product Manifold



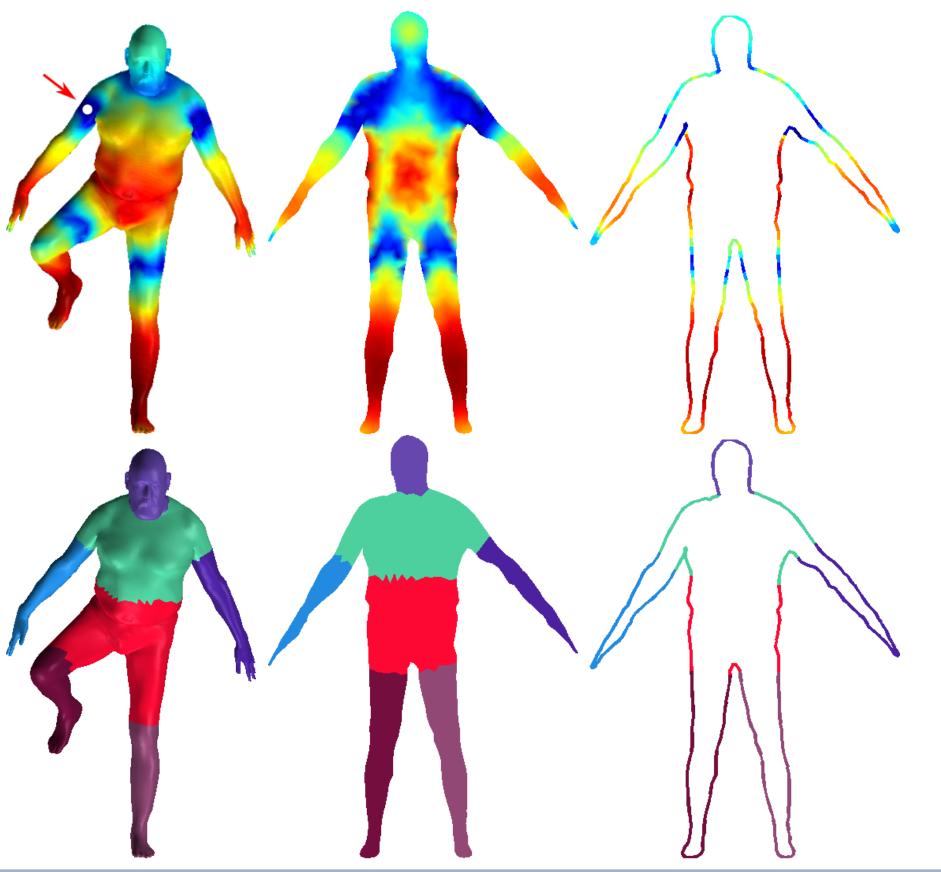


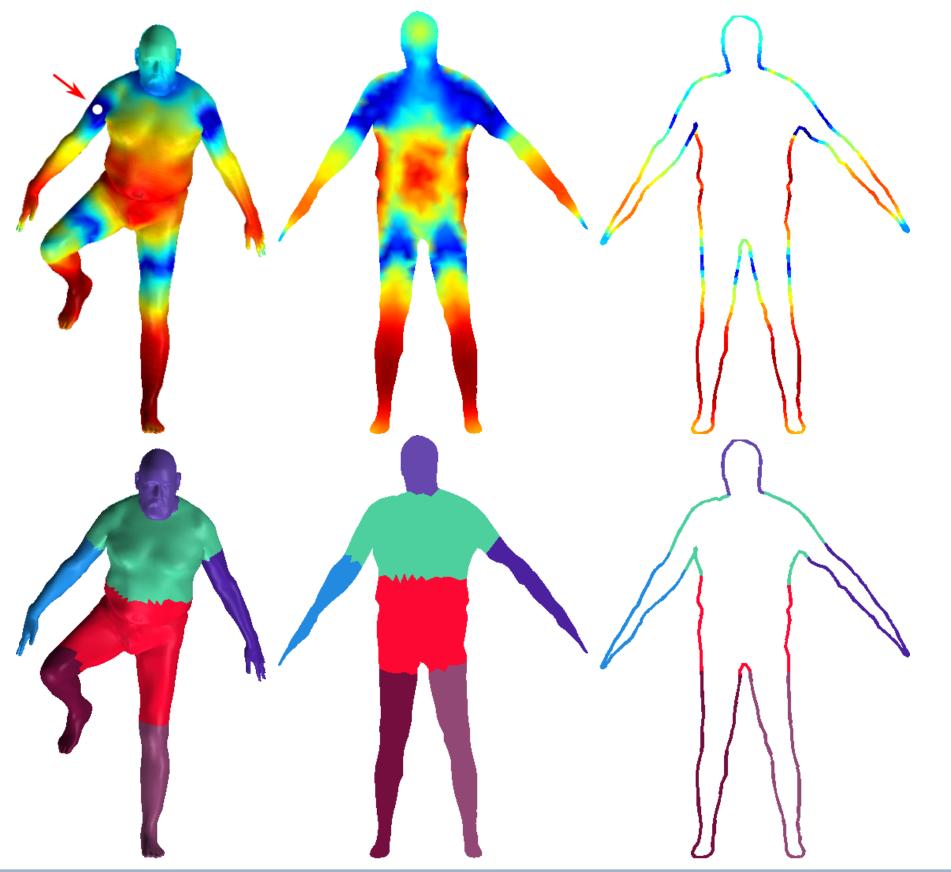
where $\Gamma_{\varphi} \subset M \times N$ denotes the graph of φ .

boils down to a **shortest path problem** on the threedimensional **product manifold** solvable by the Dijkstra algorithm [1,2] and sped up by a branch-andbound approach [3].



(d) input 3D (a) input 2D (b) Assumption: Sketches represent all discrimitative extremities of the 3D shape.





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Energy Formulation

Given two comparable features f_M , f_N we minimize the following energy:

$$E(\varphi) = \int_{\Gamma_{\varphi}} \operatorname{dist}(f_M(s_1), f_N(s_2)) d\mathbf{s}$$

 $rg\min E(\varphi)$

Spectral Features

We use the scaled HKS, the WKS and Consensus Segmentations as features.

cat huma dog horse wolf MAP centaur dog [3] [5]

(1)

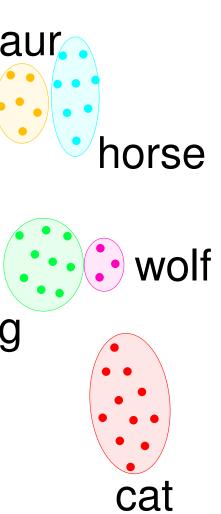
(2)



Retrieval Results

The evaluation was done on the whole TOSCA data set and templates of each null pose.

	Ours	ShapeDNA [4]	Consensus [5]
	1.0000	0.1310	0.1050
an	1.0000	0.9078	0.6532
	1.0000	0.1066	0.3330
è	0.5062	0.0611	0.0629
	1.0000	0.0379	0.0302
	0.9012	0.2489	0.2369



. 3 different humans

Embedding produced using all pairwise matching energies by tSNE.

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[2] T. Schoenemann and D. Cremers. Globally Optimal Image Segmentation with an Elastic Shape Prior. In Proc. ICCV, 2007.

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[4] M. Reuter, F.-E. Wolter and N. Peinecke. Laplace-Beltrami Spectra as 'Shape-DNA' of Surfaces and Solids. Comput. Aided Design, 2006.

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Contact

More information can be found at vision.in.tum.de/~laehner/Elastic2D3D/ (including C++ Code and the data set) Email: laehner@in.tum.de