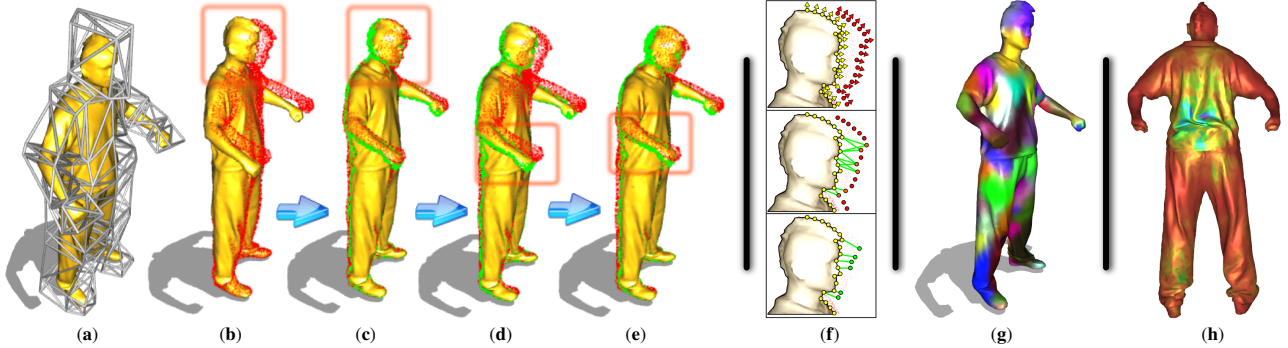


# Iterative Cage-based Registration for Dynamic Shape Capture

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**Figure 1: Life-like Dynamic Registration of Clothed Organic Shape.** A generic humanoid-type cage is enveloping a laser-scanned template mesh roughly, to perform scalable registration of whole-body scanned models (a). We aim at deforming the source template (in yellow color) toward a target point cloud (in red color) (b). The skin-detached registration procedure pulls the cage-based geometry toward corresponding target locations (c). After a sufficient number of decade iterations, the next target point cloud is reconstructed (d), and the overall iterative procedure is repeated (e). Normal-guided pairwise correspondences (in green color) are pruned, satisfying a criteria based on smoothed normals difference (f). Consequently, handle-aware overlapping harmonic-rigidities (g) are well-suited to register the non-rigid edge-length deviation (h) of the surface with controllable flexibility.

## 1 Introduction

Recent advances in low-cost dynamic scanning turn the cross-parametrization of non-rigid animatable surface into a vision-oriented ill-posed problem. In contrast with [Li et al. 2012], we propose a novel detail-preserving registration approach with resolution-independent control. Furthermore, our skin-detached surface registration avoids patch-based segmentation or affine fitting to maintain the local plasticity, as required in [Budd and Hilton 2010]. In particular, we leverage the problem of highly non-rigid spacetime registration by employing an elasto-plastic coarse cage. Thus, we perform scalable handle-aware harmonic shape registration, relying on the high-level of shape abstraction offered by the space-based paradigm. To the best of our knowledge, our technique is the first to investigate handle-aware elastic overlapping-rigidities for registering life-like dynamic shapes in full-body clothing.

## 2 Handle-Aware Detached Registration

**Non-Rigid Registration Setup.** We propose to evolve the fixed connectivity  $\mathcal{F}$ , offered by the given dense template mesh  $\mathcal{M} = \{\mathcal{V}, \mathcal{F}\}$ , by registering roughly-and-temporally a given set  $\mathcal{P} = \{\mathcal{P}_0, \dots, \mathcal{P}_l\}$  of  $l$  unstructured time-varying point clouds. We assume no prior knowledge about the temporal matching. The geometry of the template mesh is written by  $\mathcal{V}$ . Let us denote  $\Omega \subset \mathbb{R}^3$  the bounded domain included by  $m$  control cage-handles enveloping the static template mesh. We designate by  $\mathbf{c}_j$  the current location of the  $j^{\text{th}}$  cage handle in the global coordinates system. This cage polytope structure is augmented by the Laplace-Beltrami Operator  $\mathcal{L}(\cdot)$  with non-uniform cotangent weights, allowing scalable template mesh registration. Associated differential cage coordinates  $\delta$  encode each cage-handle relatively to its neighborhood in the cage connectivity. Finally, the geodesic-aware relationship between the volumetric subspace and the static template is encapsulated by a bi-harmonic rigging process, computed once at the default pose.

**Normal-Guided Pairwise Correspondences.** We adopt a similar fuzzy-yet-robust geometric strategy than [Budd and Hilton 2010] to infer a minimal set of compatible feature correspondences  $\mathcal{S} = \{s_k : (k, \mathbf{q}_k, \gamma_k)\}$ , updated at each intra-frame iteration. A unique target location  $\mathbf{q}_k \in \mathbb{R}^3$  is obtained for the  $k^{\text{th}}$  current template vertex by averaging candidates in the current point cloud of  $\mathcal{P}$ . After the outliers pruning, each correspondence is weighted by  $\gamma_k$  defined as the dot product of smoothed pairwise normals.

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**Iterative Elasto-Plastic Optimization.** For each point cloud, our cage-handle curve registration process alternates successively between deformation optimization and correspondences. At each intra-frame iteration  $t+$ , we initialize the cage geometry with location  $\mathbf{c}^t$  obtained at the previous iteration  $t$ . Then, we update  $\mathcal{L}(\cdot)$ , the corresponding  $\delta$  and we infer  $\mathcal{S}$ . Finally, driven by the correspondences propagation in the subspace, a new cage pose  $\mathbf{c}^{t+}$  is estimated by solving the following variational objective function:

$$\operatorname{argmin}_{\{\mathbf{c}_1^{t+}, \dots, \mathbf{c}_m^{t+}\}} \left( \alpha \cdot \sum_{j=1}^m \|\mathcal{L}^t(\mathbf{c}_j^{t+}) - \delta_j^t\|_2^2 + \beta \cdot \sum_{s_k \in \mathcal{S}^t} \gamma_k \cdot \left\| \mathbf{q}_k - \sum_{j=1}^m w_{kj} \cdot \mathbf{c}_j^{t+} \right\|_2^2 \right)$$

where  $w_{kj} : \Omega \rightarrow \mathbb{R}$  is the biharmonic weight for a given cage handle  $j$  with respect to the  $k^{\text{th}}$  template vertex, as proposed in [Jacobson et al. 2011]. Consequently, the registered template geometry is generated by a cage-based warping field with low-distortion.

**Weight-Control Update Rules.** The data-term weight-control is initialized at  $\beta = 0.01$  and increases along iterations by following an exponential growth rule to promote the constraint-guided bending energy. The weight-control  $\alpha$  enforcing the shape-prior is set-up to 1 and slightly decreases to relax the deformation stiffness prior.

## 3 Conclusions

Our new approach is a first step toward the automatic template-based registration of highly non-rigid dynamic shape using low-dimensional space-based encoding. We train the effectiveness of our algorithm by aligning several real-world datasets of [Vlasic et al. 2008]. The main advantage of our iterative optimization remains in the simultaneous cross-reconstruction of dynamic shape, and skin-detached registration of reusable temporal curves expressing the clothed-body deformations. In brief, we proposed a new system that registers shape variations while preserves the life-likeness of captured data, and acquires reusable consistent surface parameters. We expect to pursue our on-going efforts to perform better fine-tuned non-rigid alignment for large organic motion.

## References

- BUDD, C., AND HILTON, A. 2010. Temporal alignment of 3d video sequences using shape and appearance. In *CVMP'10: 9th Conference on Visual Media Production*.
- JACOBSON, A., BARAN, I., POPOVIĆ, J., AND SORKINE, O. 2011. Bounded biharmonic weights for real-time deformation. *ACM Trans. Graph.* 30.
- LI, H., LUO, L., VLASIC, D., PEERS, P., POPOVIĆ, J., PAULY, M., AND RUSINKIEWICZ, S. 2012. Temporally coherent completion of dynamic shapes. *ACM Trans. Graph.* 31.
- VLASIC, D., BARAN, I., MATUSIK, W., AND POPOVIĆ, J. 2008. Articulated mesh animation from multi-view silhouettes. *ACM Trans. Graph.* 27 (August), 97:1–97:9.