

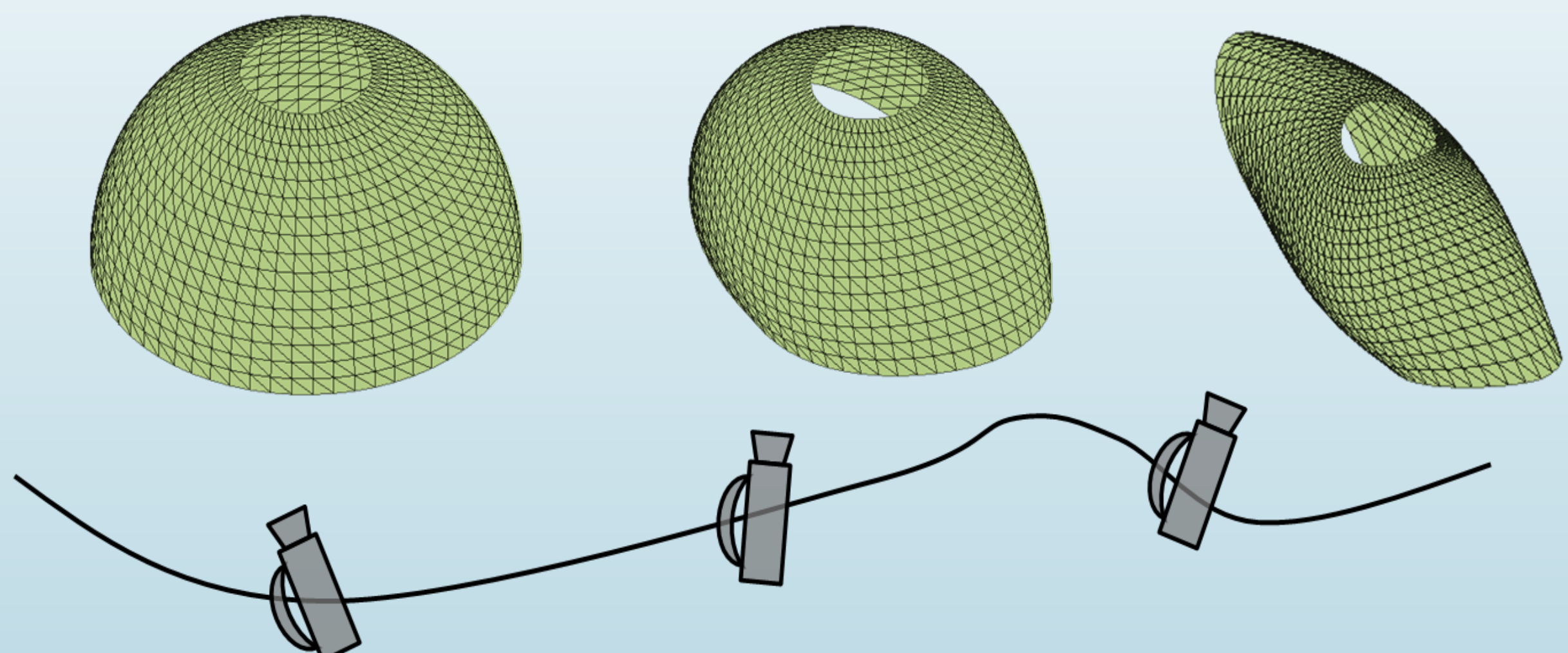


GOOD VIBRATIONS: A MODAL ANALYSIS APPROACH FOR SEQUENTIAL NON-RIGID STRUCTURE FROM MOTION

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NON-RIGID SfM

- 3D reconstruction of non-rigid objects from 2D temporal correspondences in a monocular image sequence.
- So far most approaches are batch.
- **Our Goal:** A sequential NRSfM method that is **real-time capable**.



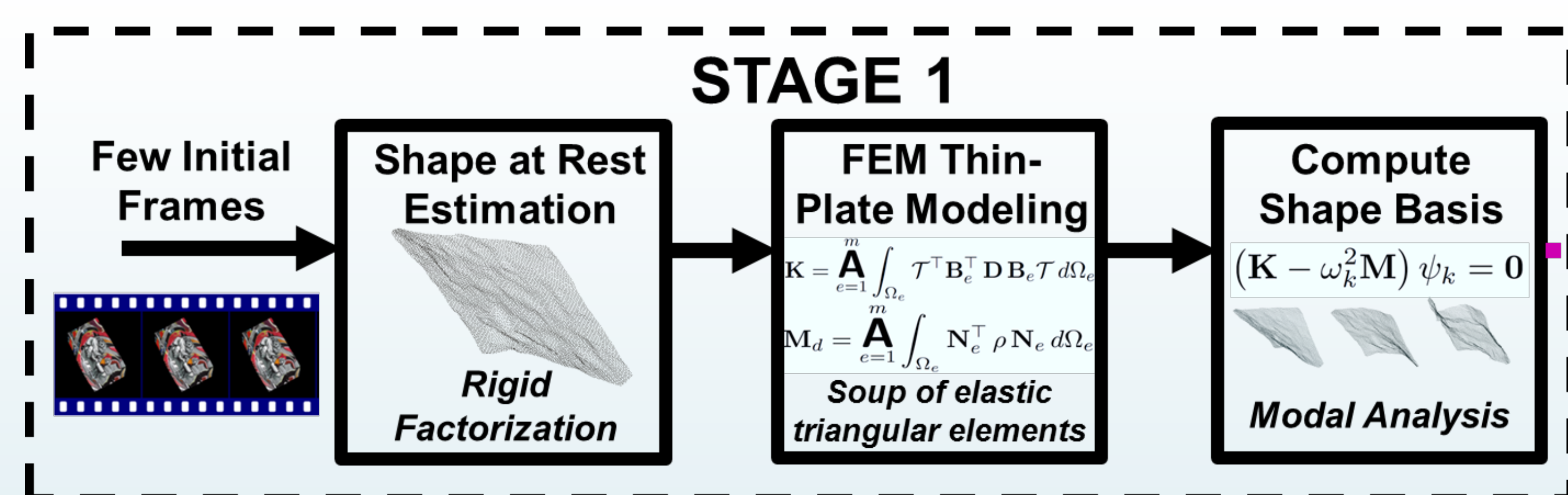
OUR CONTRIBUTION

- A **new shape basis** based on modal analysis with bending and stretching modes. No learning step.
- An **online solution** to NRSfM that estimates camera pose and deformable shape on a per-frame basis.

OUR APPROACH

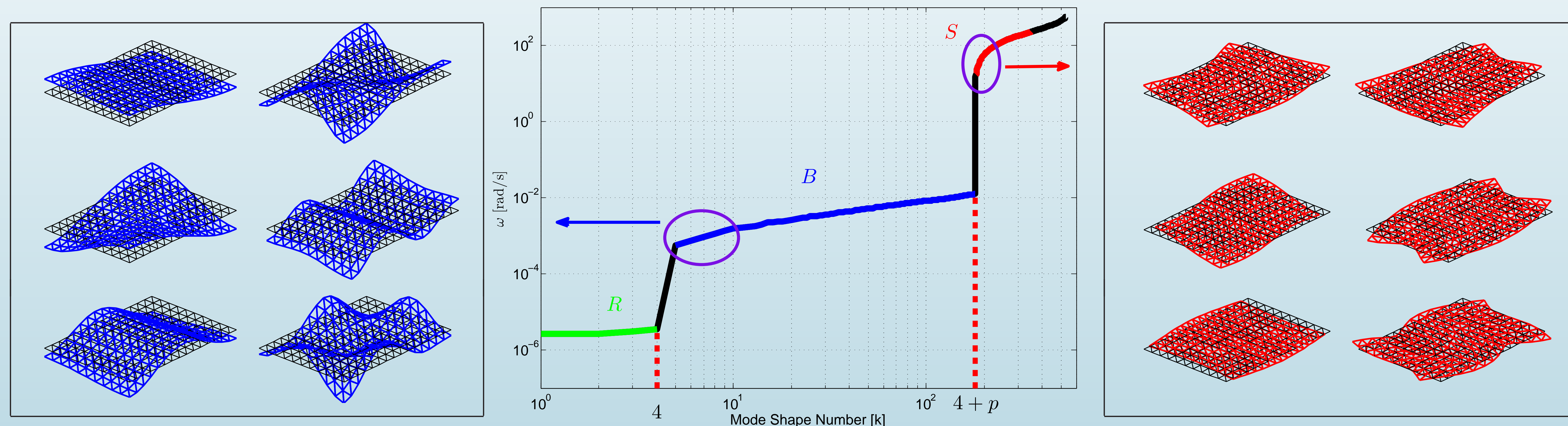
- **Stage 1:** Computation of the shape basis. Only an estimation of the 3D shape at rest is needed.
- **Stage 2:** Bundle Adjustment over a sliding window to optimize deformation coefficients and rigid pose.
- Suitable to model a wide variety of deformations: from inextensible to highly **extensible surfaces**.

STAGE 1: PHYSICS-BASED MODAL SHAPE BASIS COMPUTATION



- Mode shapes are ordered by *frequency spectrum*, i.e. according to the energy needed to excite each mode:

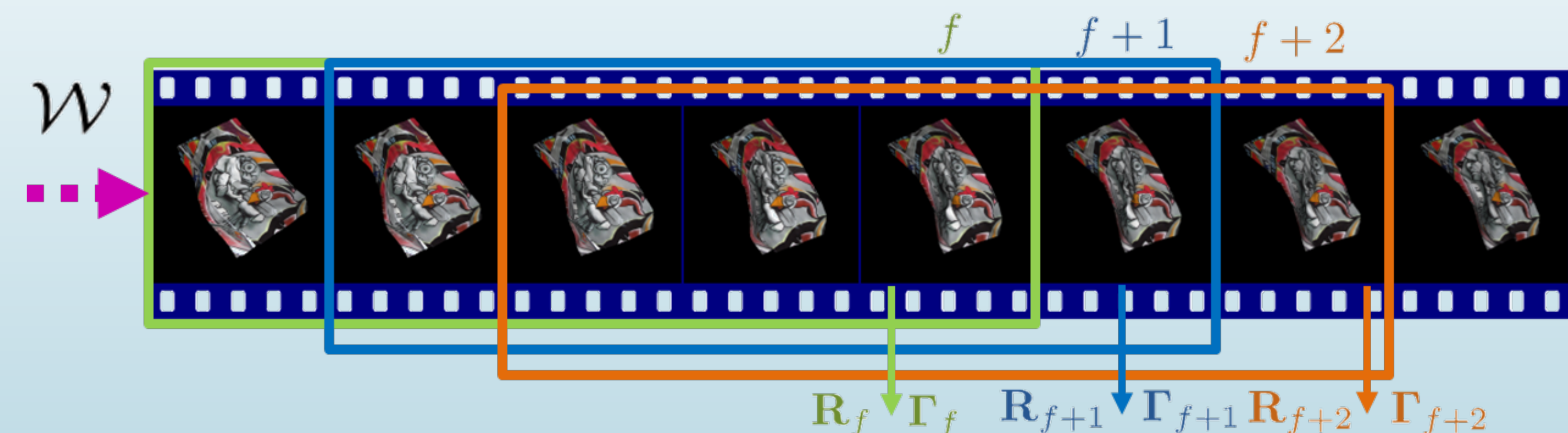
- **Rigid modes (R):** Rigid body motions.
- **Bending modes (B):** out-of-plane deformations. Interval $[5, p+4]$ (*quasi-isometric*).
- **Stretching modes (S):** in-plane extensible deformations. Interval $[p+5, 2p]$.



STAGE 2: ONLINE SLIDING WINDOW OPTIMIZATION

Our energy includes image reprojection error terms for all visible points within a temporal sliding window of W frames in addition to temporal smoothness priors. The optimization is solved using sparse Levenberg-Marquardt:

$$\arg \min_{\mathbf{R}_i, \Gamma_i} \sum_{i=f-W+1}^f \sum_{\rho \in \mathcal{M}} \|\mathbf{W}_{i\rho} - \mathbf{R}_i (\mathbf{S}_\rho + (\mathbf{I}_3 \otimes \Gamma_i^T) \Psi_\rho)\|_{\mathcal{F}}^2 + \lambda_q \sum_{i=f-W+1}^f \|\mathbf{q}_i - \mathbf{q}_{i-1}\|_{\mathcal{F}}^2 + \lambda_\gamma \sum_{i=f-W+1}^f \|\Gamma_i^T - \Gamma_{i-1}^T\|_{\mathcal{F}}^2$$



- Non-rigid 3D displacement is approximated as a **linear combination** with Γ_i weight coefficients.
- Orthographic camera model:

$$\mathbf{W}_f = \begin{bmatrix} u_{f1} & u_{f2} & \dots & u_{fp} \\ v_{f1} & v_{f2} & \dots & v_{fp} \end{bmatrix} = \Pi \mathbf{Q}_f (\mathbf{S}_f + \mathbf{T}_f)$$

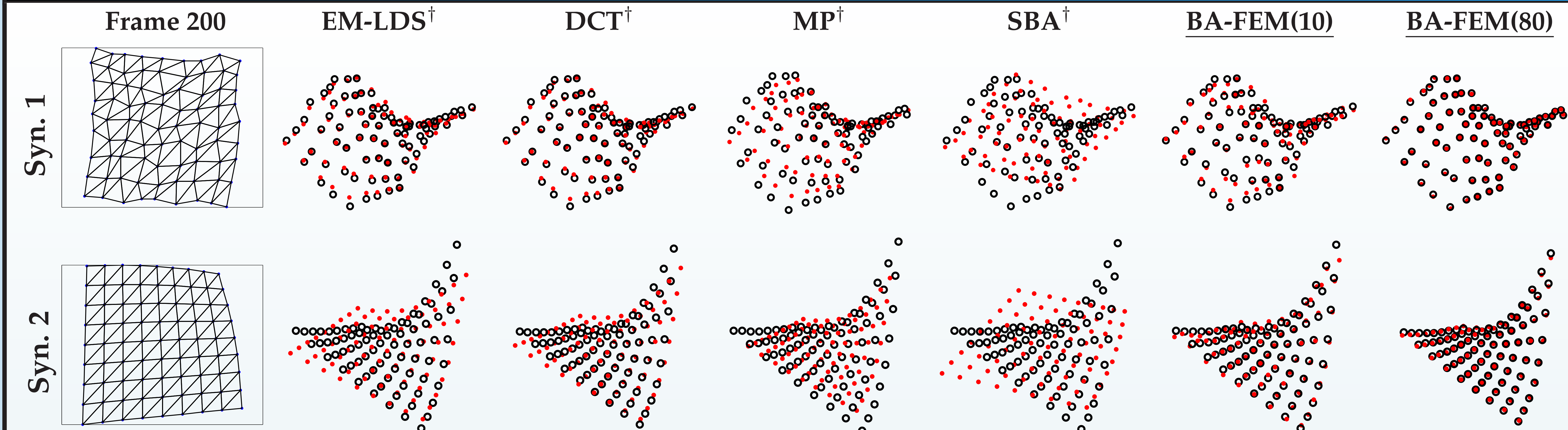
CONCLUSIONS

- Real-time capable, accurate and scalable solution.
- Our method can handle both sparse and dense meshes and can cope with structured missing data.
- No additional inextensibility constraints.

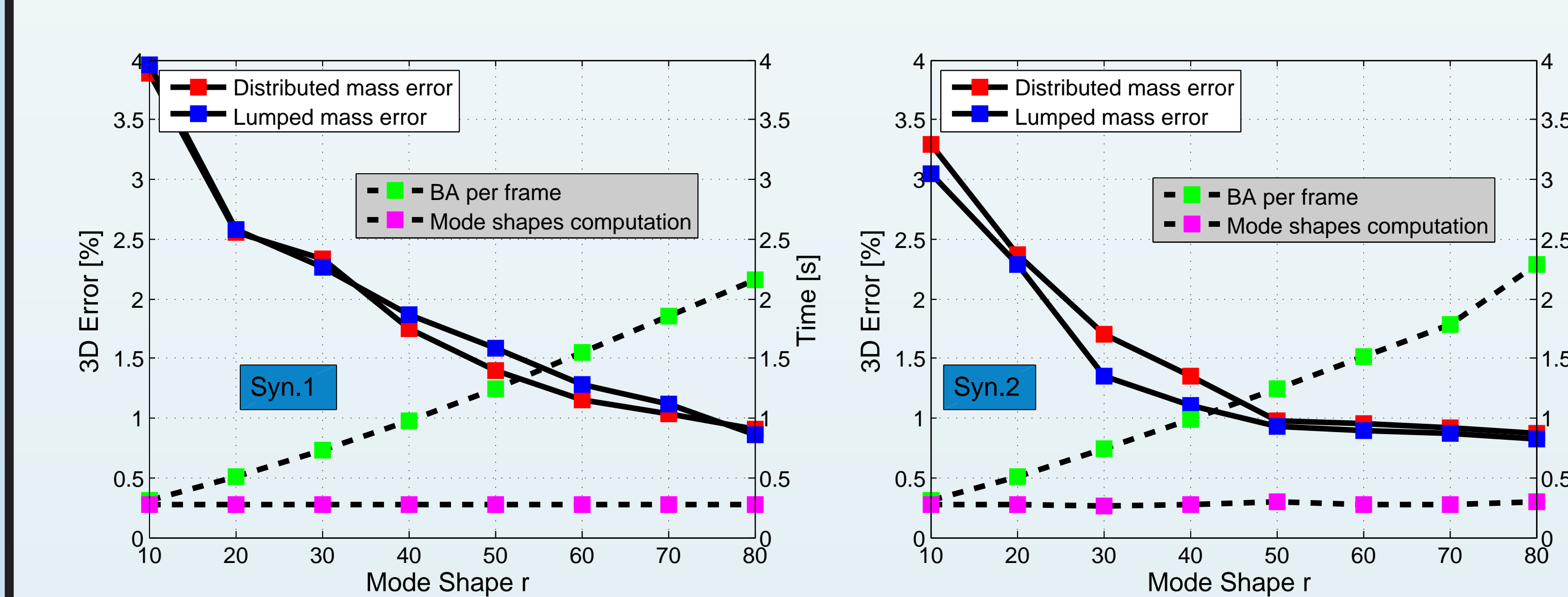
FUTURE WORK

- Incorporate feature tracking and outlier detection to provide a unified framework.
- To exploit biomechanical priors.
- Application to Minimally Invasive Surgery.

EXPERIMENTAL RESULTS



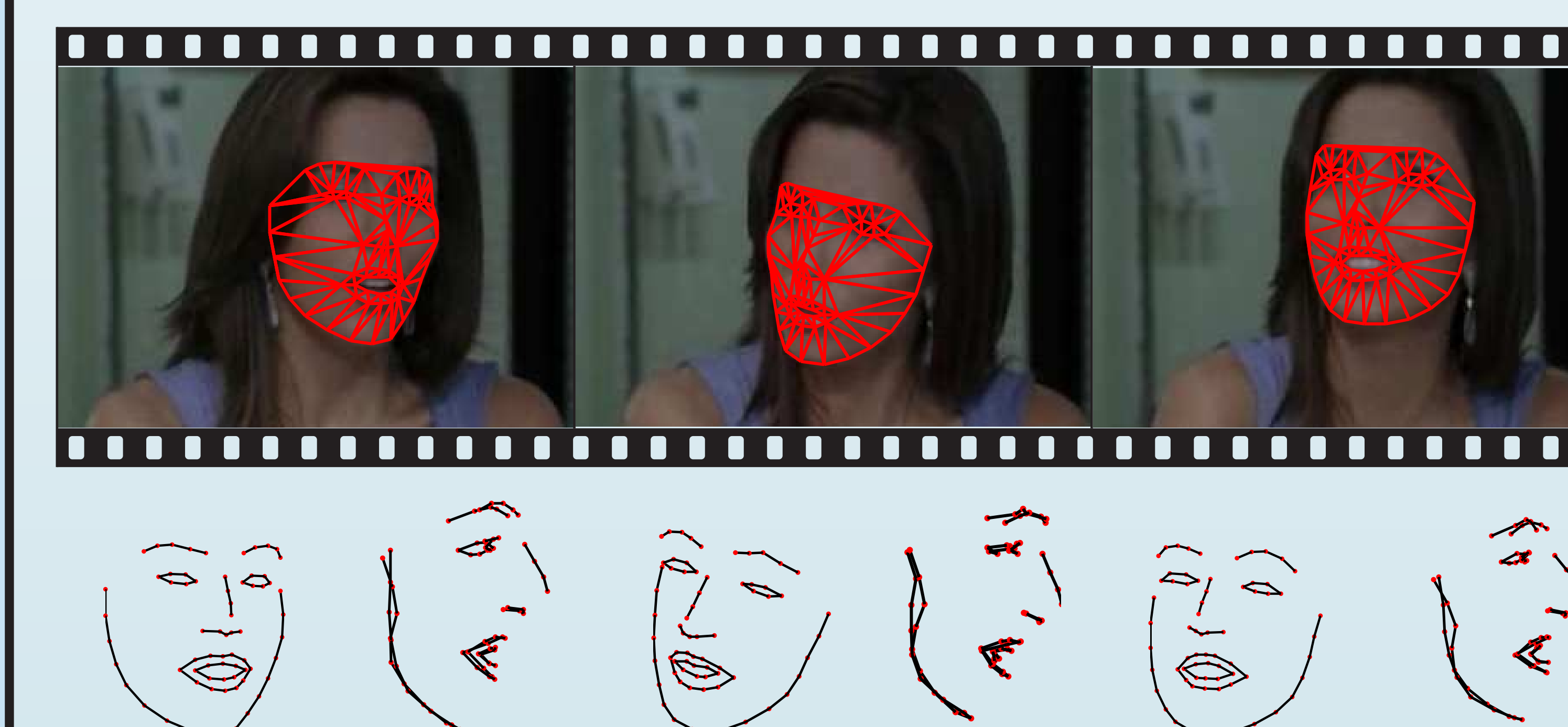
SYNTHETIC SEQUENCES



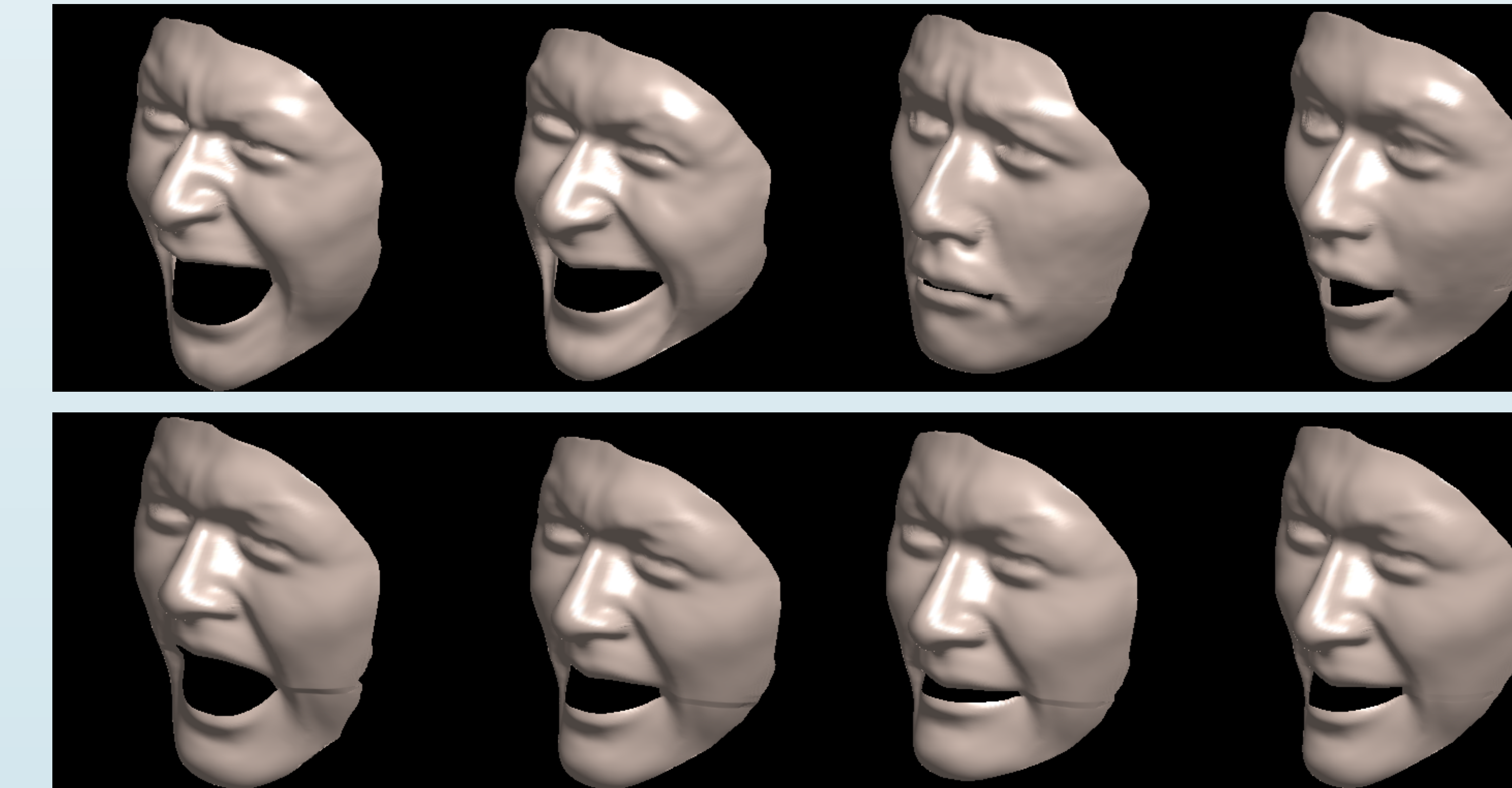
$e_{3D}\%$	Syn. 1	Syn. 2	B [†]	S [†]
EM-LDS [†]	11.12(2)	10.92(2)	✓	
DCT [†]	9.25(2)	11.81(5)	✓	
MP [†]	12.42(2)	18.84(2)	✓	
SBA [†]	14.03(16)	20.90(8)		✓
BA-FEM	3.89(10)	3.04(10)		✓
BA-FEM	0.86(80)	0.82(80)		✓

[†] EM-LDS [Torresani et al. PAMI'08], DCT [Akhter et al. PAMI'11], MP [Paladini et al. CVPR'09], SBA [Paladini et al. ECCV'10]. B: Batch algorithm and S: Sequential algorithm.

ACTRESS SEQUENCE



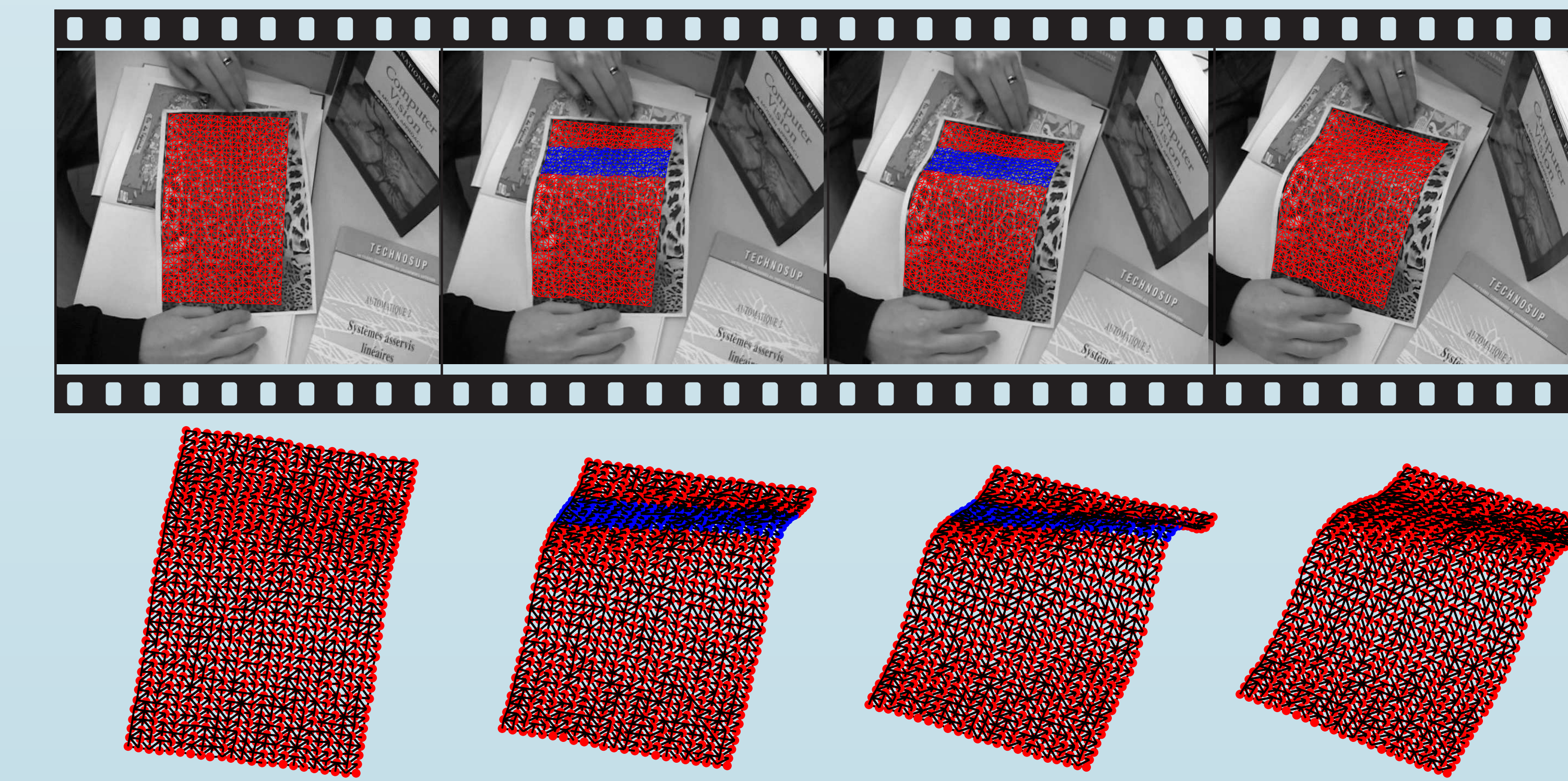
DENSE FACE SEQUENCE



DENSE FLAG SEQUENCE



PAPER BENDING SEQUENCE with STRUCTURED OCCLUSION



Seq.	Size		Run-time (sec)		Error $e_{3D}\%$
	p^\ddagger	r^\ddagger	MSC [†]	BA [†]	
Syn. 1	81	10-80	0.3	0.3-2.2	3.89-0.86
Syn. 2	81	10-80	0.3	0.3-2.2	3.04-0.82
Actress	68	10	0.4	0.3	-
Bending	828	10	60	0.85	-
Flag	9,622	5-25	300	47-416	4.14-3.29
Face	28,887	5	1540	60	4.64

[†] MSC: Modes Shape Computation, BA: Bundle Adjustment time, p and r number of points and modes respectively.

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