

SIMULTANEOUS POSE AND NON-RIGID SHAPE WITH PARTICLE DYNAMICS

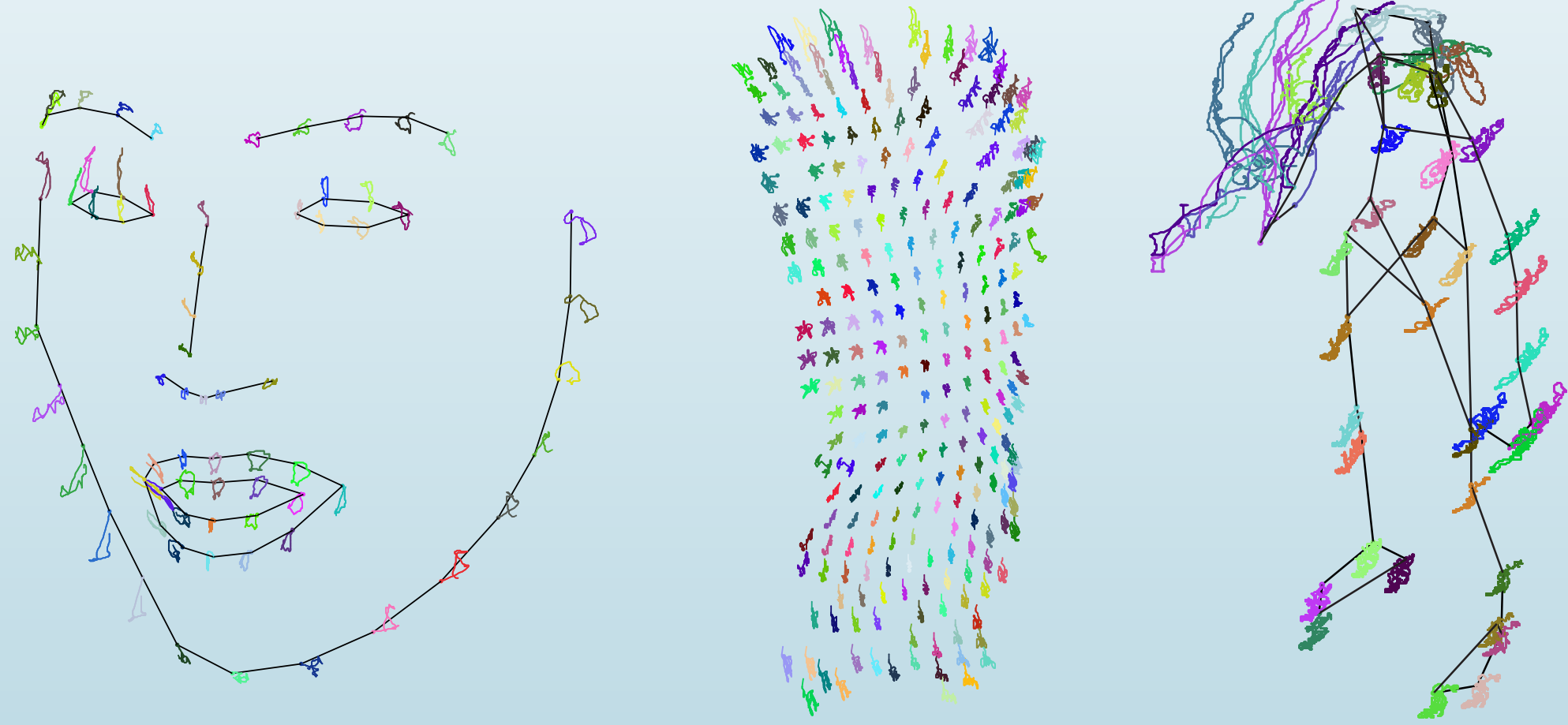
ANTONIO AGUDO¹, FRANCESC MORENO-NOGUER²

¹I3A-UNIVERSIDAD DE ZARAGOZA, ²INSTITUT DE ROBÒTICA I INFORMÀTICA INDUSTRIAL (CSIC-UPC)



PROBLEM: NON-RIGID SFM

- 3D reconstruction of deformable objects from 2D temporal tracks in a monocular video.
- So far most approaches use global models and batch operation.
- *Our Goal:* A *sequential* NRSfM method based on *local* models that is **real-time** capable.



CONTRIBUTION

- A piecewise model at **particle level** based on physical constraints to model the non-rigid deformation between consecutive frames.
- An **online solution** to NRSfM that estimates time-varying shape and camera motion.

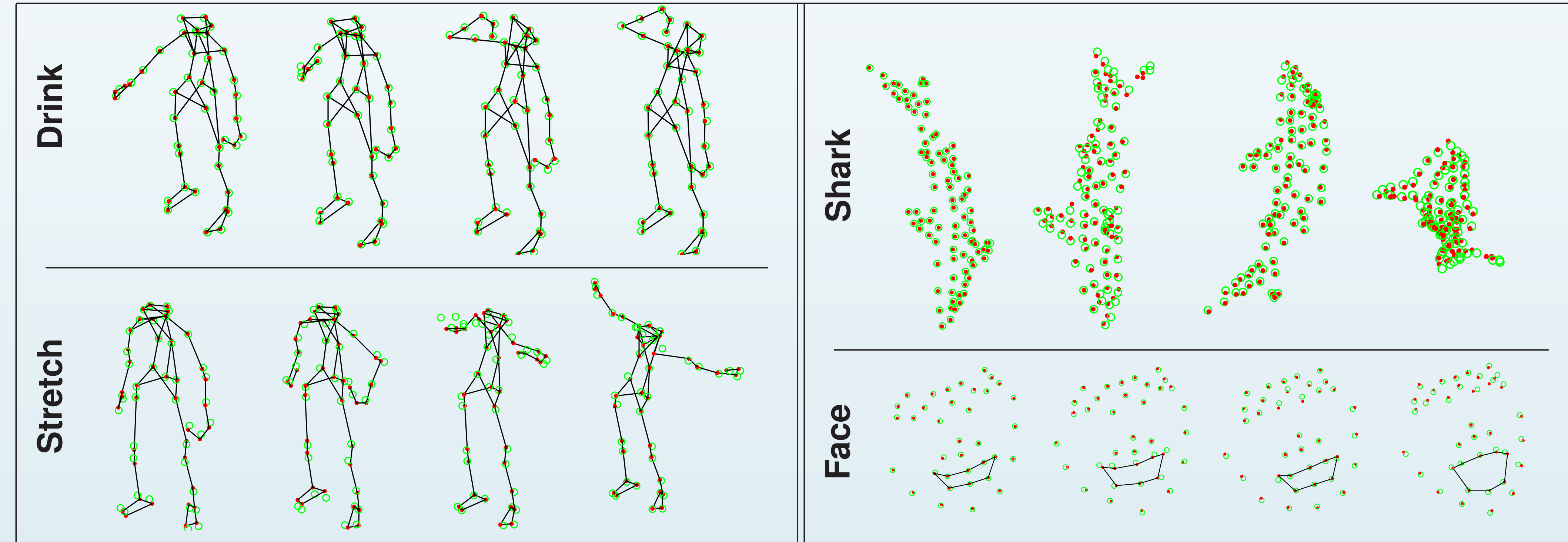
APPROACH

- A dynamic object is modeled as a system of particles with force-perturbed motion models.
- Model parameters are estimated by Bundle Adjustment over a sliding window of images.
- Suitable to model a wide variety of deformations: from *articulated* to *non-rigid* motion even for *discontinuous* surfaces.

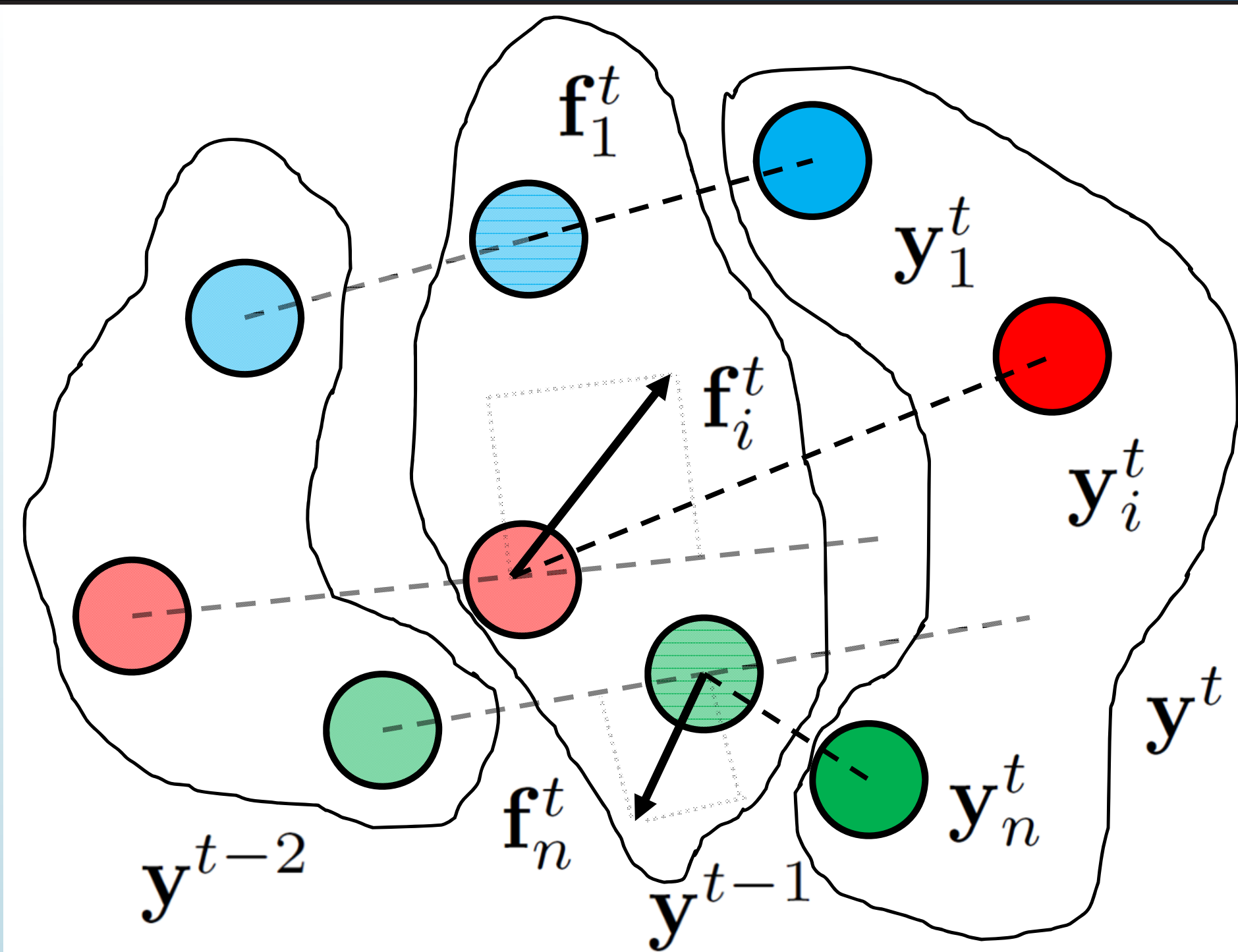
EXPERIMENTAL EVALUATION

Seq.	Met.	Batch Methods					Sequential Methods			
		EM-PPCA [†]	MP [†]	PTA [†]	CSF2 [†]	KSTA [†]	SPM [†]	SBA [†]	BAFEM [†]	PSMM
Drink [‡]		5.56(5)	4.14(6)	1.38(13)	1.14(6)	0.94(12)	1.60(12)	11.25(12)	-	1.93
Stretch [‡]		13.72(15)	8.13(5)	3.85(8)	2.46(8)	2.00(7)	1.86(11)	17.61(20)	-	5.76
Yoga [‡]		11.89(14)	12.98(8)	2.42(8)	1.84(7)	2.12(7)	1.65(10)	15.84(20)	-	6.65
Shark ⁺		1.82(2)	9.34(23)	5.91(6)	1.09(5)	1.03(3)	6.29(2)	8.81(5)	-	6.99
Jacky ⁺		1.80(5)	2.74(5)	2.69(3)	1.93(5)	2.12(4)	1.82(7)	2.90(16)	3.43(15)	2.80
Face ⁺		7.30(9)	3.77(7)	5.79(2)	6.34(5)	6.14(8)	2.67(9)	6.92(27)	6.89(2)	4.49

[†] EM-PPCA [Torresani et al. PAMI'08], MP [Paladini et al. CVPR'09], PTA [Akhter et al. PAMI'11], CSF2 [Gotardo et al. CVPR'11], KSTA [Gotardo et al. ICCV'11], SPM [Dai et al. CVPR'12], SBA [Paladini et al. ECCV'10], BA-FEM [Agudo et al. CVPR'14]. [‡] Articulated motion. ⁺ Non-rigid motion.



PHYSICS-INSPIRED MOTION MODEL



- We employ *Newton's second law of motion* $f_i^t = m_i a_i^t$ to constrain deformation per point. The acceleration at time t is approximated using second-order finite differences:

$$f_i^t \approx m_i \left[\frac{y_i^{t-2} - 2y_i^{t-1} + y_i^t}{(\Delta t)^2} \right]$$

- The 3D position of the particles at time t follows the dynamical model:

$$y^t = f^t + 2y^{t-1} - y^{t-2} = f^t + d^t$$

- The constant velocity model d^t is force-perturbed by f^t at each image.

ONLINE NON-LINEAR OPTIMIZATION

- Orthographic camera model:

$$P^t = [p_1^t, \dots, p_n^t] = R^t Y^t + T^t$$

The global energy we define includes *image reprojection* error terms for all visible points within a temporal sliding window of 3 frames in addition to *spatial* and *temporal* smoothness priors. The optimization is solved using sparse Levenberg-Marquardt:

$$A(R^j, t^j, F^t) = \sum_{j=t-2}^t \sum_{v \in V^j} \|p_v^j - R^j y_v^j - t^j\|_{\mathcal{F}}^2 + \alpha_e \sum_{e=1}^{n_e} \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{d_e^2}{2\sigma^2}\right) |d_e^t - d_e^t(F^t)| + \alpha_s \|Y^t(F^t) - Y^{t-1}\|_{\mathcal{F}}^2 + \alpha_p \sum_{j=t-1}^t \|q^j - q^{j-1}\|_{\mathcal{F}}^2 + \alpha_t \sum_{j=t-1}^t \|t^j - t^{j-1}\|_{\mathcal{F}}^2$$

- We model the non-rigid deformation of an object represented by a system of particles.

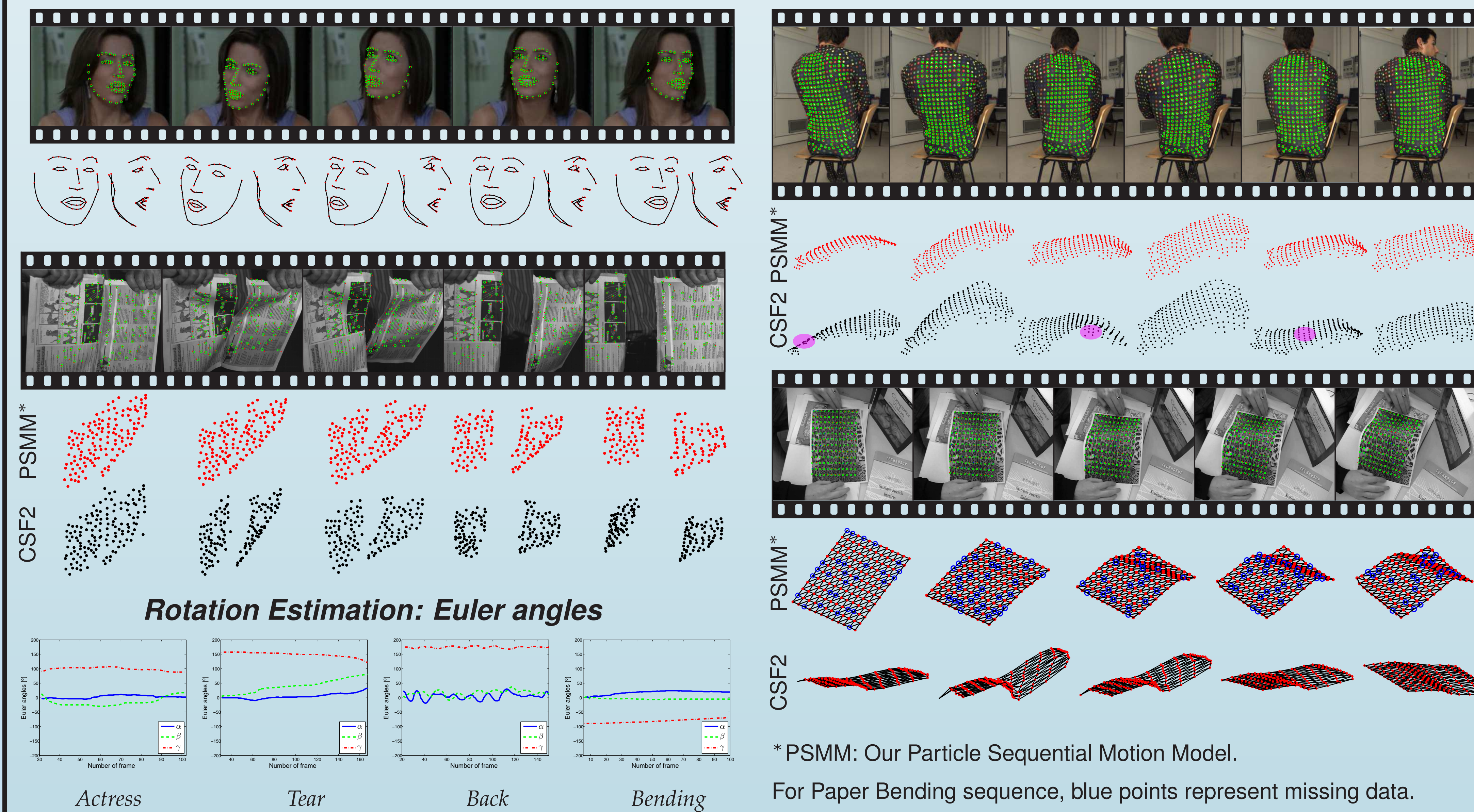
CONCLUSIONS

- Newton's second law of motion to model non-rigid deformations into bundle adjustment.
- Our method can handle different types of deformations and it can cope with missing data.

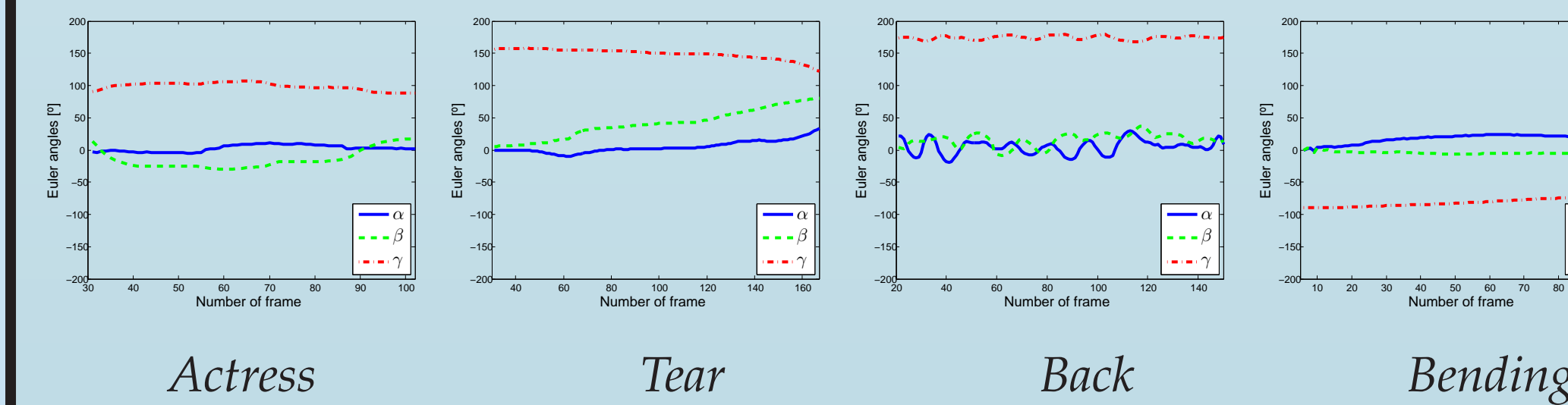
FUTURE WORK

- Generalization to full perspective cameras.
- To cope multiple non-rigid objects.
- Simultaneous feature tracking and outliers detection into a single process.

Real Video Sequences



Rotation Estimation: Euler angles



*PSMM: Our Particle Sequential Motion Model. For Paper Bending sequence, blue points represent missing data.