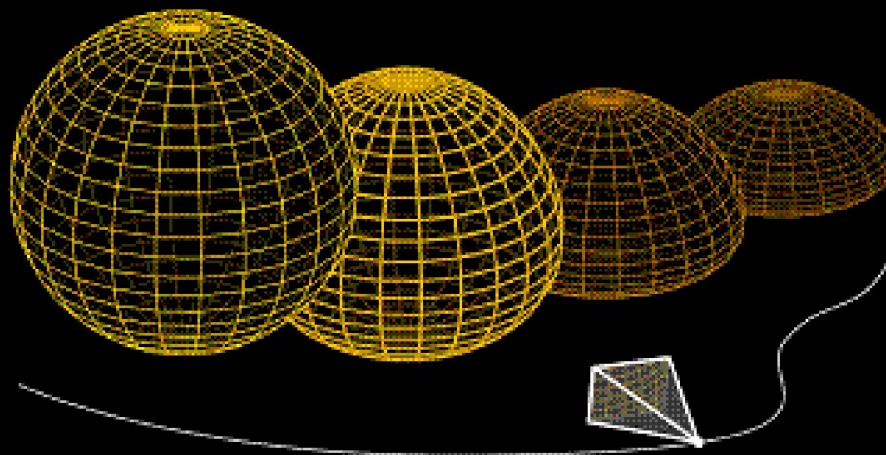
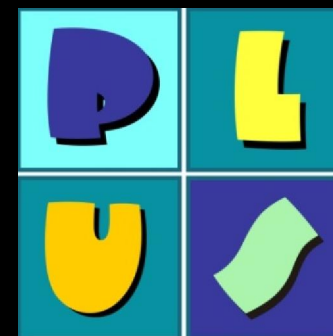


Adaptive metric registration of 3D Models to Non-rigid image trajectories



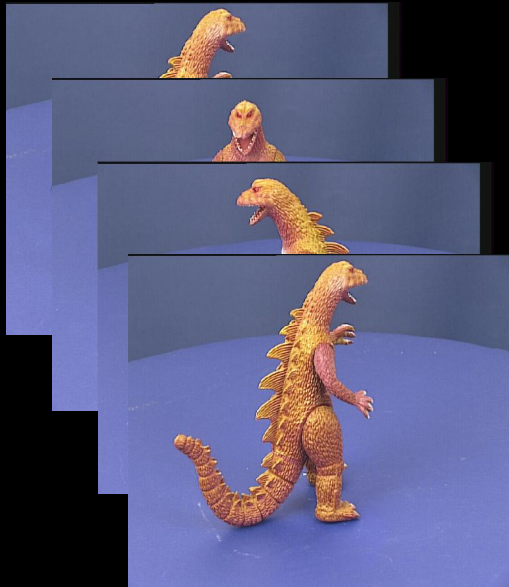
Alessio Del Bue
PLUS LAB
Italian Institute of Technology



ECCV 2010, Crete, 7th of September, 2010

Image Sequence Registration

R
i
g
i
d



=

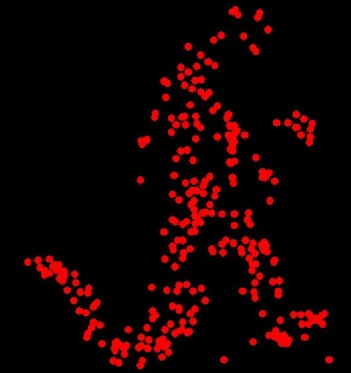


R

+

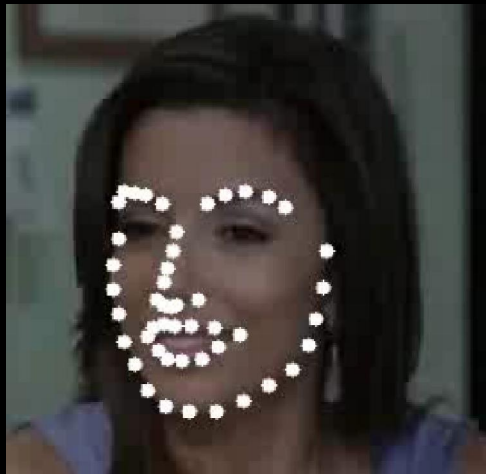
T

Unknowns



3D shape

A
d
a
p
t
i
v
e



=

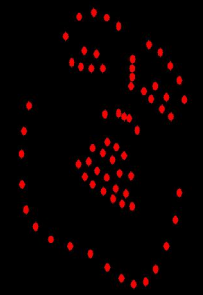


R

+

T

Unknowns



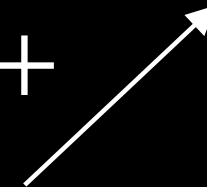
Rigid Registration

$$\begin{bmatrix} u_{11} & \dots & u_{1P} \\ v_{11} & \dots & v_{1P} \\ u_{21} & \dots & u_{2P} \\ v_{21} & \dots & v_{2P} \\ u_{51} & \dots & u_{5P} \\ v_{51} & \dots & v_{5P} \\ u_{41} & \dots & u_{4P} \\ v_{41} & \dots & v_{4P} \end{bmatrix}$$

=



+



R

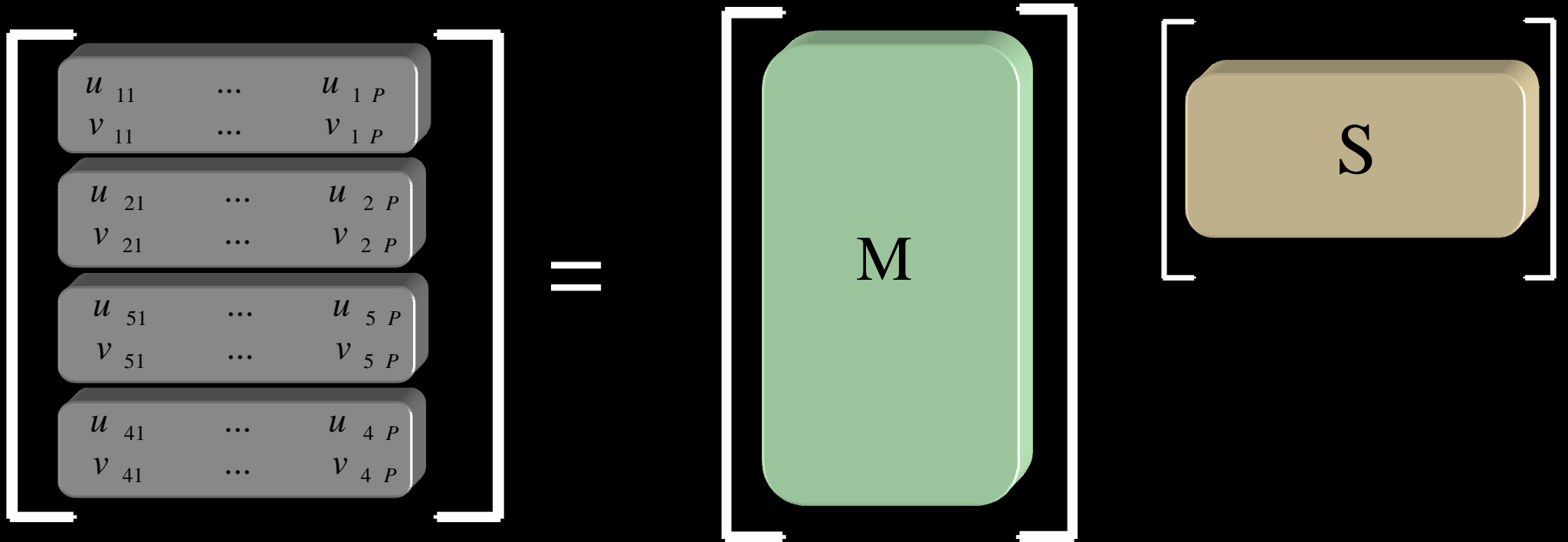
T

Unknowns

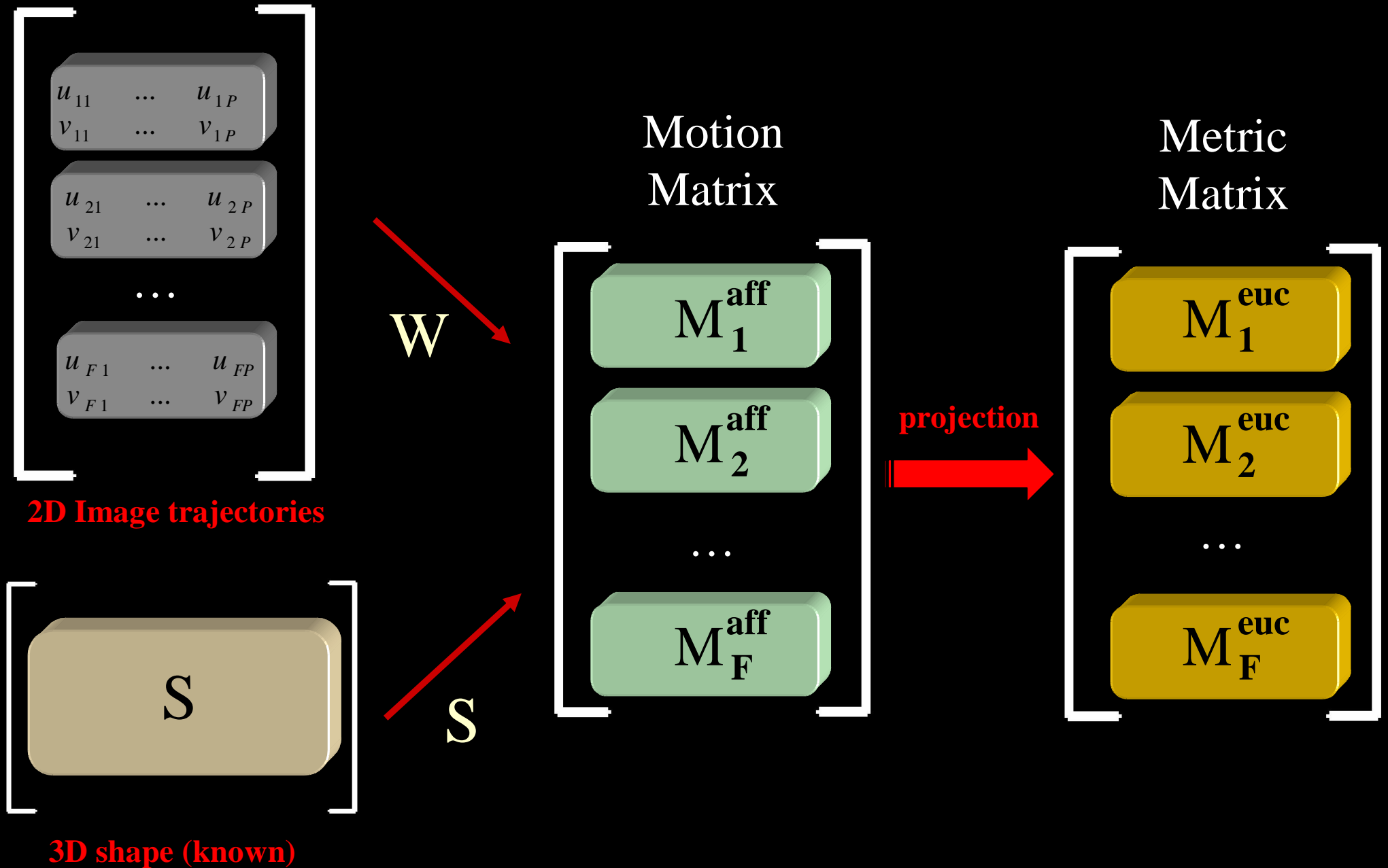


3D shape

Rigid Registration

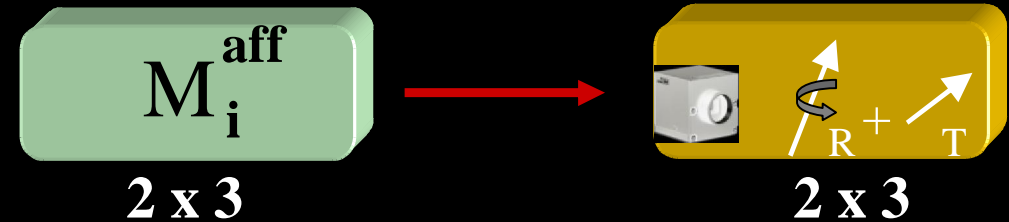


1. Rigid Registration - Affine



2. Rigid Registration – Metric

Problem: Given an initial solution find the matrix which respects the given metric constraints.



- The constraints on N_i depends on the given camera model, in this case we consider the simplest **orthographic camera** model:

$$M_i^{\text{euc}} \quad M_i^{\text{euc}}{}^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

2×3 3×2 2×2

- The projection to the closest orthographic camera matrix can be solved optimally via SVD

$$M_i^{\text{aff}} \xrightarrow{\text{SVD (econ)}} U \quad D \quad V^T$$

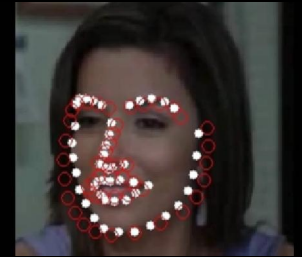
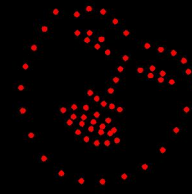
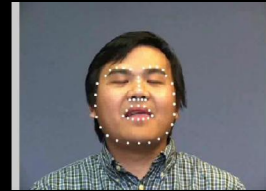
2×3 2×2 2×2 2×3

$$M_i^{\text{euc}} = U \quad V^T$$

2×3 2×2 2×3

Adaptive Metric Registration

Problem: Now the 3D shape is not an exact description of the observed 2D image trajectories. Thus we need to estimate a **new shape** that adapts to the measured 2D data.



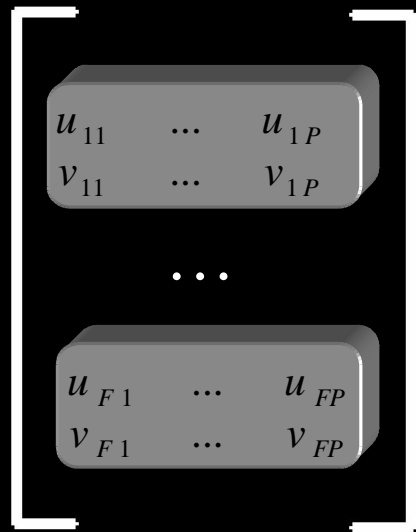
3D model building

registration

Proposed Solution: A two step closed form procedure.

1. Find a **common bilinear decomposition** for the 2D image trajectories and the shape to register
2. Find the best transformation which enforces the **metric constraints** for both the 2D trajectories and the 3D shape to register

1. Find a joint decomposition



2D Image trajectories

$$\xrightarrow{\text{GSVD}} \mathbf{U} \mathbf{D}_U \mathbf{X}^T$$



3D shape (known)

$$\xrightarrow{\text{GSVD}} \mathbf{V} \mathbf{D}_V \mathbf{X}^T$$

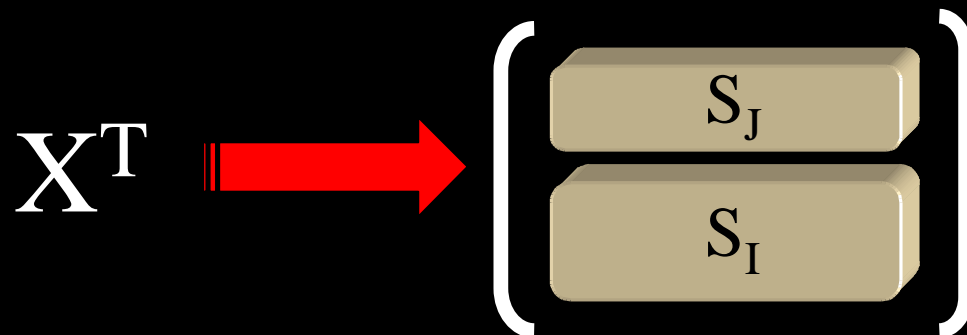
- Common row space \mathbf{X}^T given the image trajectory and 3D shape

- Generalised singular values such that:

$$\mathbf{D}_u^2 + \mathbf{D}_v^2 = \mathbf{I}$$

- Orthogonal column space such that

$$\mathbf{U} \mathbf{U}^T = \mathbf{I} \text{ and } \mathbf{V} \mathbf{V}^T = \mathbf{I}$$



After simple matrix manipulations it is possible to extract an affine joint shape given both data.

2. Adaptive Registration – Metric

$$\mathbf{W} = \begin{bmatrix} \mathbf{M}_J & \mathbf{M}_I \end{bmatrix} \begin{bmatrix} \mathbf{S}_J \\ \mathbf{S}_I \end{bmatrix}$$

$$\mathbf{S} = \begin{bmatrix} \mathbf{N}_J \end{bmatrix} \begin{bmatrix} \mathbf{S}_J \end{bmatrix}$$

- The solution from GSVD is affine, force the **metric constraints** given the camera matrix.

- There is an **infinite set** of solutions for the joint factorisation:

$$\mathbf{W}_j = \mathbf{M}_j \mathbf{S}_j \quad \text{and} \quad \mathbf{S} = \mathbf{N}_j \mathbf{S}_j$$

- Since any full rank matrix $\mathbf{Q}_{3 \times 3}$ gives:
 $\mathbf{W}_j = \mathbf{M}_j \mathbf{Q} \mathbf{Q}^{-1} \mathbf{S}_j \quad \text{and} \quad \mathbf{S} = \mathbf{N}_j \mathbf{Q} \mathbf{Q}^{-1} \mathbf{S}_j$

$$\mathbf{W}_j = \tilde{\mathbf{M}}_j \tilde{\mathbf{S}}_j \quad \text{and} \quad \mathbf{S} = \tilde{\mathbf{N}}_j \tilde{\mathbf{S}}_j$$

- Find the **corrective transform** \mathbf{Q} that enforces the metric constraints **both for the 2D trajectories and the known 3D shape**.

- We propose a new set of constraints given the 3D **metric constraints** in \mathbf{S} i.e. \mathbf{N}_j should represent a rotation.

2. Joint Metric Constraints

$$\begin{array}{ccc}
 \begin{array}{c} M_i \\ 2 \times 3 \end{array} & \begin{array}{c} M_i^T \\ 3 \times 2 \end{array} & = \begin{array}{c} \begin{bmatrix} a & c \\ c & b \end{bmatrix} \\ 2 \times 2 \end{array} \\
 \begin{array}{c} M_i^{euc} \\ 2 \times 3 \end{array} & \begin{array}{c} M_i^{euc T} \\ 3 \times 2 \end{array} & = \begin{array}{c} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \\ 2 \times 2 \end{array} \\
 \end{array}
 \quad \text{and} \quad
 \begin{array}{ccc}
 \begin{array}{c} N_J \\ 3 \times 3 \end{array} & \begin{array}{c} N_J^T \\ 3 \times 3 \end{array} & = \begin{array}{c} C \\ 3 \times 3 \end{array} \\
 \begin{array}{c} N_J^{euc} \\ 3 \times 3 \end{array} & \begin{array}{c} N_J^{euc T} \\ 3 \times 3 \end{array} & = \begin{array}{c} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ 3 \times 3 \end{array}
 \end{array}$$

Find the corrective transform Q such that:

$$\begin{array}{ccc}
 \begin{array}{c} M_i \\ 2 \times 3 \end{array} & \begin{array}{c} Q \\ 3 \times 3 \end{array} & \begin{array}{c} Q^T \\ 3 \times 3 \end{array} & \begin{array}{c} M_i^T \\ 3 \times 2 \end{array} & = & \begin{array}{c} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \\ 2 \times 2 \end{array} \\
 \begin{array}{c} N_J \\ 3 \times 3 \end{array} & \begin{array}{c} Q \\ 3 \times 3 \end{array} & \begin{array}{c} Q^T \\ 3 \times 3 \end{array} & \begin{array}{c} N_J^T \\ 3 \times 3 \end{array} & = & \begin{array}{c} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ 3 \times 3 \end{array}
 \end{array}$$

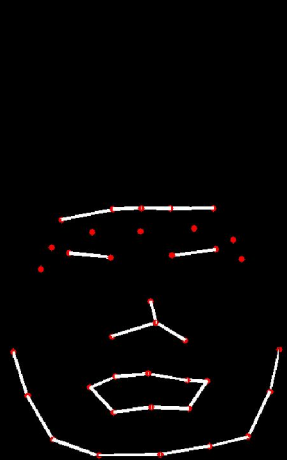
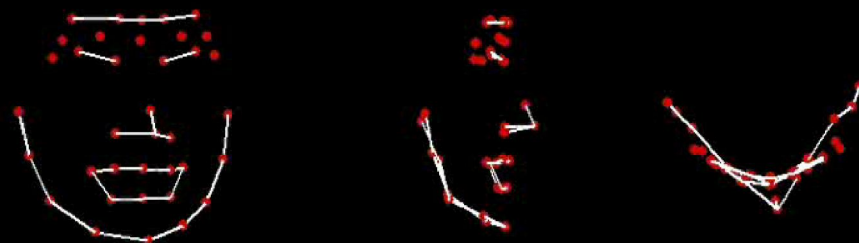
Two solutions for Q are proposed: one with Least Squares and the other with Convex optimisation (in the paper).

Experiments

Synthetic Experiments



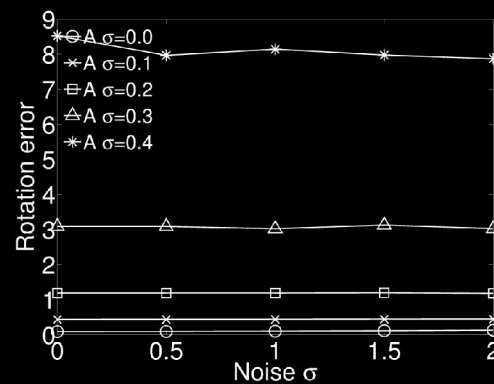
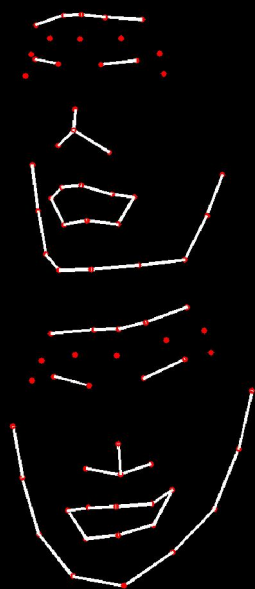
Motion capture
ground truth



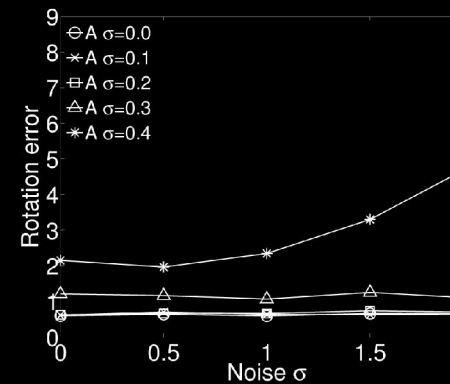
3D shape



Apply
Distortion

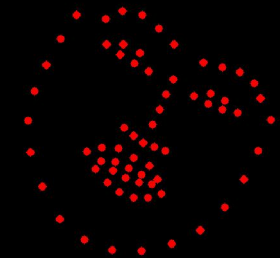
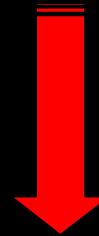
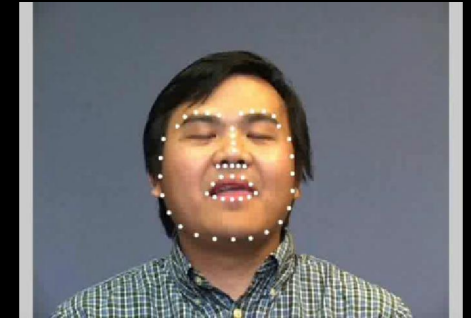
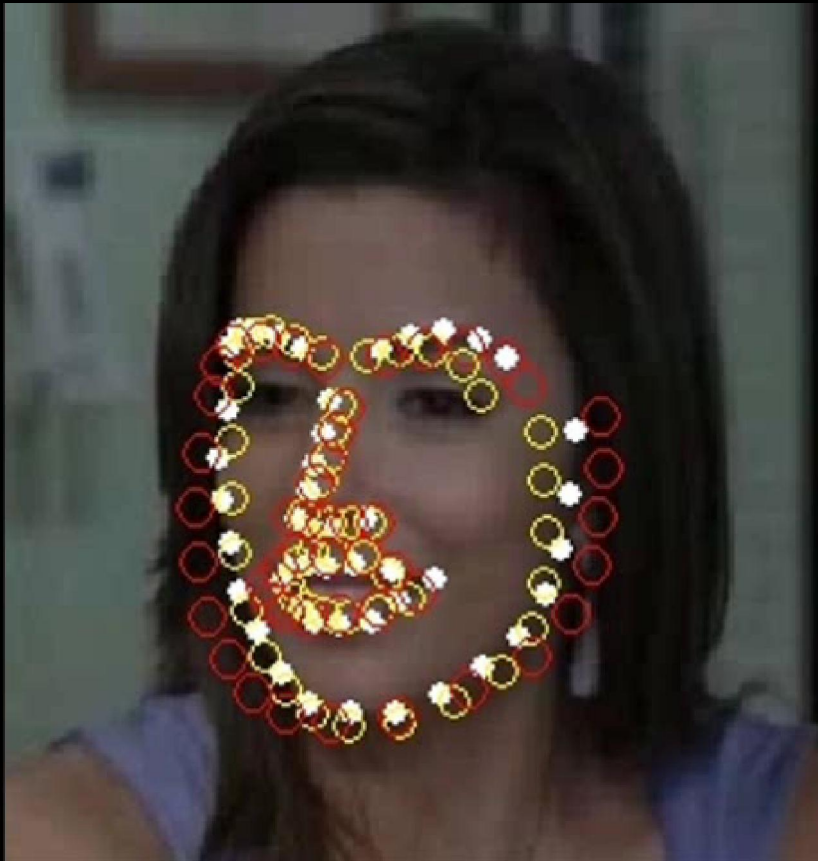


Rigid



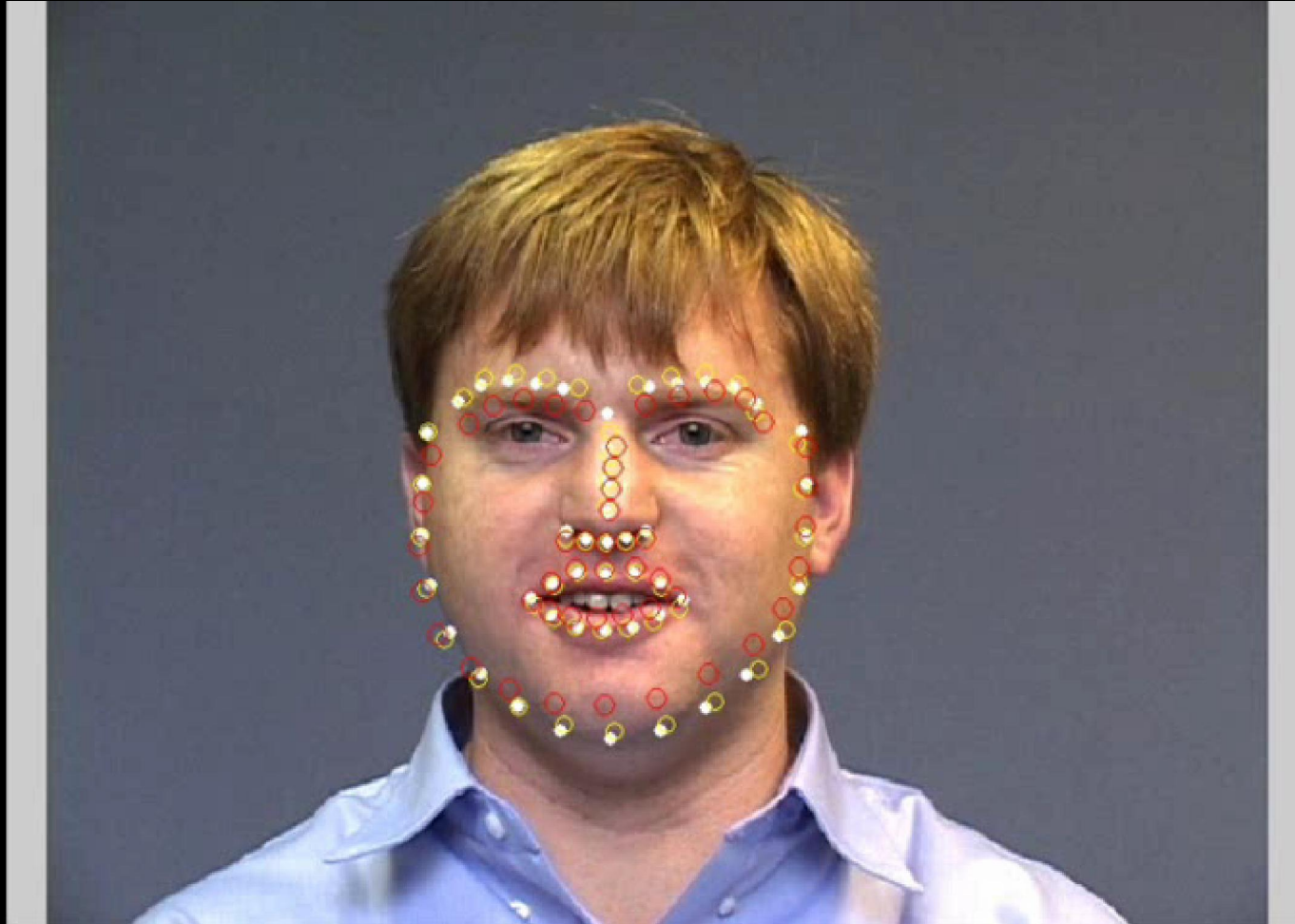
Adaptive

Real Experiments



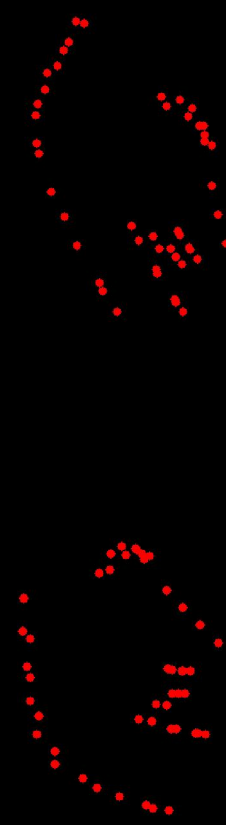
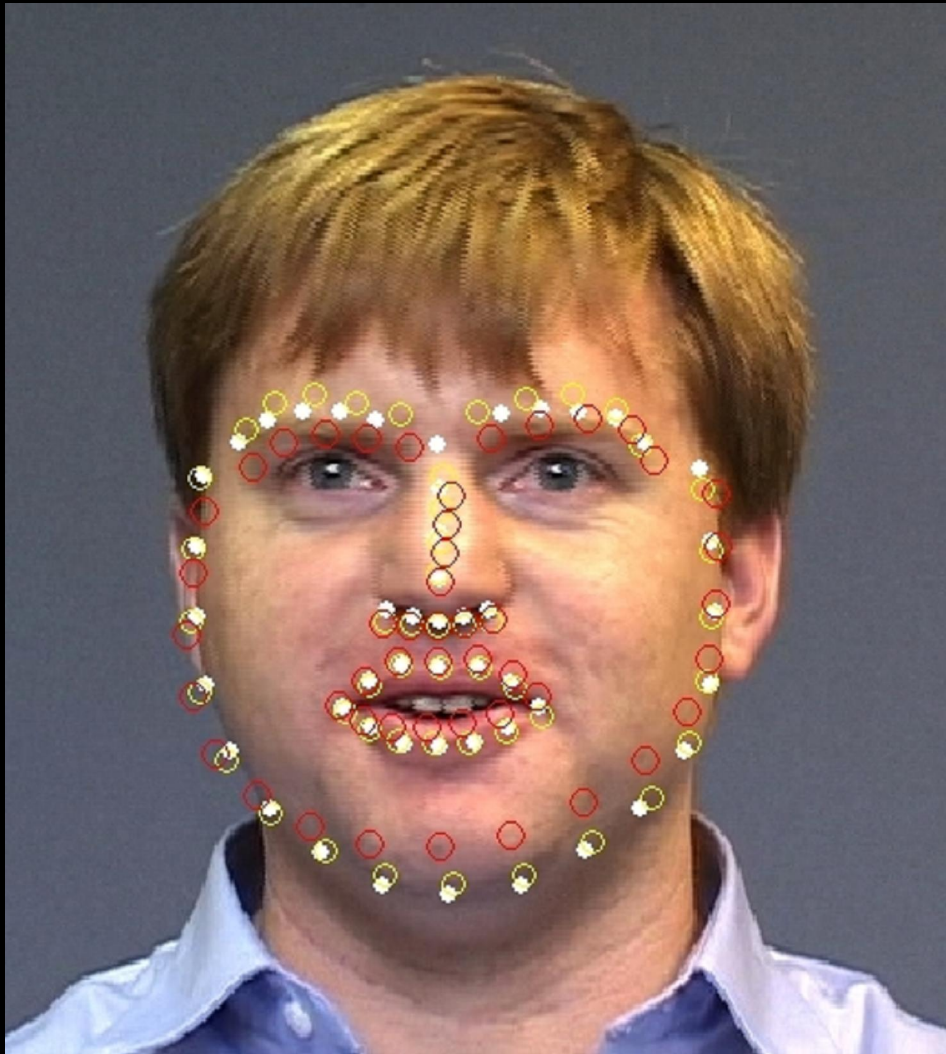
- White dots represent the tracked 2D data (Tracks and images courtesy of Julien Peyras)
- Red circles show the registration **without adaptation**.
- Yellow circles show the registration **with adaptation**.

Real Experiments



- White dots represent the tracked 2D data (Tracks and images courtesy of Jing Xiao)
- Red circles show the registration **without adaptation**.
- Yellow circles show the registration **with adaptation**.

Real Experiments



Shape without joint metric constraints

$$R_i R_i^T = I$$

~~$$Z_i Z_j^T = I$$~~

Shape using joint metric constraints

$$R_i R_i^T = I$$

$$Z_j Z_j^T = I$$

- White dots represent the tracked 2D data (Tracks and images courtesy of Jing Xiao)
- Red circles show the registration **without adaptation**.
- Yellow circles show the registration **with adaptation**.

Real Experiments – Missing Data



Missing data algorithm:

0. Find an **initialization** for the missing entries in the measurement matrix W
 1. Compute the **joint image registration** and find M_j and S_j .
 2. Given this solution, **inpute** the missing entries as $W = M_j S_j$.
 3. Go back to 1. until **convergence** is reached.

- White dots represent the tracked 2D data (Tracks and images courtesy of Julien Peyras)
- Yellow circles show the registration **with adaptation**.

Conclusions

To summarise:

Adaptive metric registration is made by a **two step closed-form** procedure:

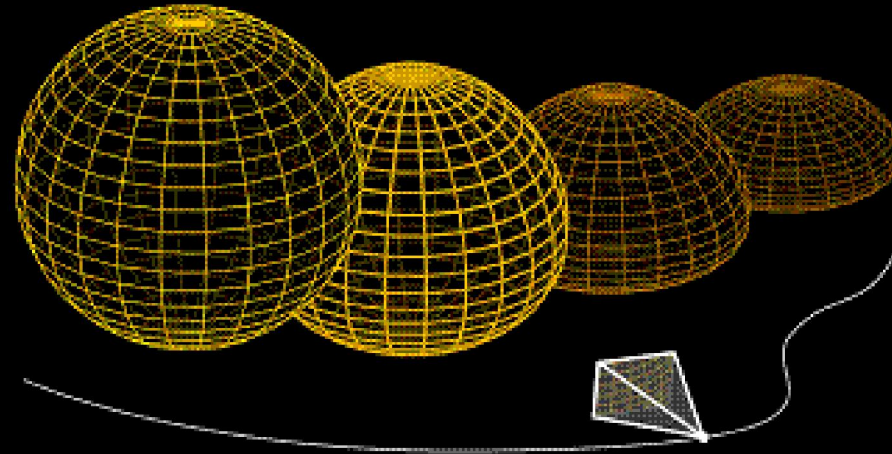
1. Compute a **common bilinear decomposition with GSVD** for the 2D image trajectories and the shape to register
2. Find the best transformation which enforces the **metric constraints** for both the 2D trajectories and the 3D shape to register

Further work:

The proposed framework can be extended to any data that can be expressed in **bilinear form** i.e. photometric stereo, non-rigid and articulated SfM.

Extend the method to **deal with outliers** – when the 3D shape and 2D data have mismatches GSVD may not find a common decomposition.

Adaptive metric registration of 3D Models to Non-rigid image trajectories



For more information:
www.isr.ist.utl.pt/~adb/

