

# Registration of 3D Faces

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CS6101 AY2012-13 Semester 1

# Main Paper

- ◎ T. J. Hutton, B. F. Buxton and P. Hammond.  
Automated Registration of 3D Faces using  
Dense Surface Models.  
In *Proc. British Machine Vision Conference*,  
2003.

# Motivation

- ◎ 3D face model useful for many applications:
  - animation
  - motion tracking
  - face recognition
  - face reconstruction
  - surgery planning & simulation
  - forensic reconstruction
  - ...

# Motivation

- ◎ Build 3D face model from training samples:



- ◎ Need to **align** them: **registration**.

# Motivation

- ⦿ Can't just align spatially:



Everything is messed up!

Need to align nose to nose, eyes to eyes, ...

# Motivation

Two general kinds of registration:

- ⦿ Rigid registration

- Objects differ by scale, rotation, translation.
- No change in shape during registration.
- Easy to solve.

- ⦿ Non-rigid registration

- Objects differ by scale, rotation, translation, shape.
- Must change shape during registration.
- Harder to solve.

# Motivation

- One possibility: manually mark **landmark** points.

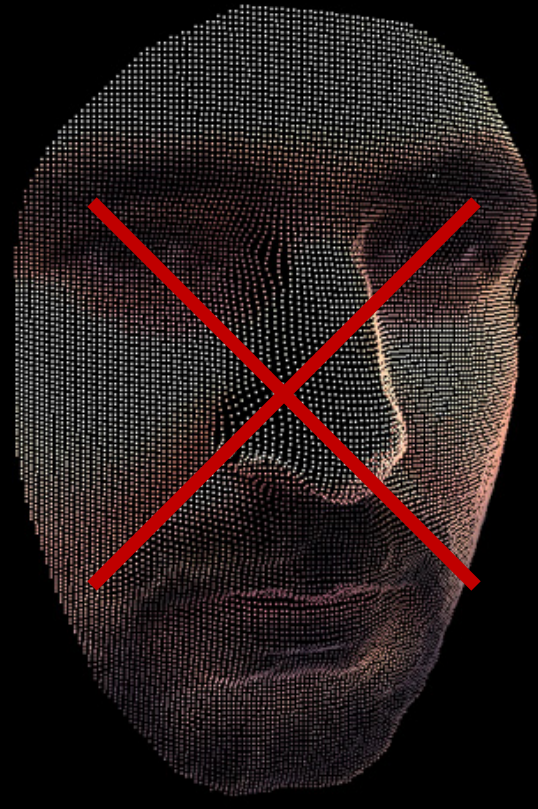
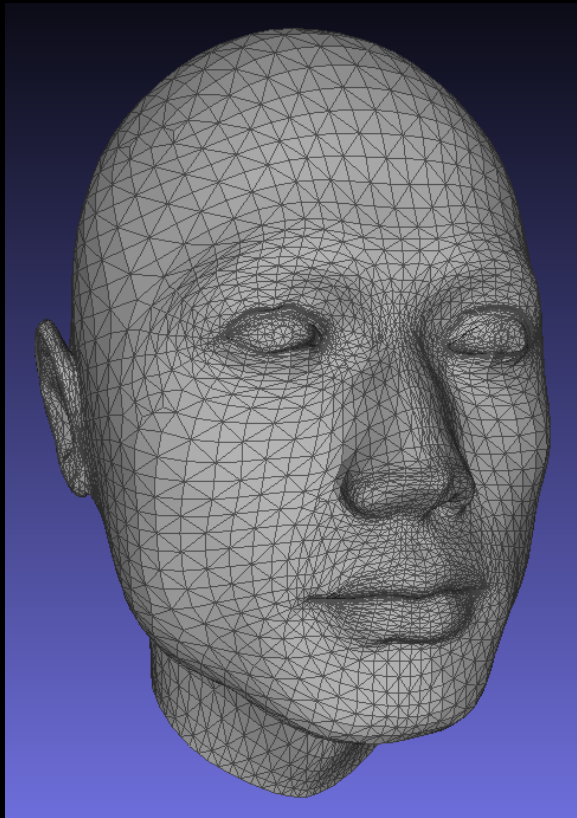


Very tedious and time-consuming!

Need automatic method!

# Focus

- 3D model has shape and texture.
- Focus on shape, leave out texture



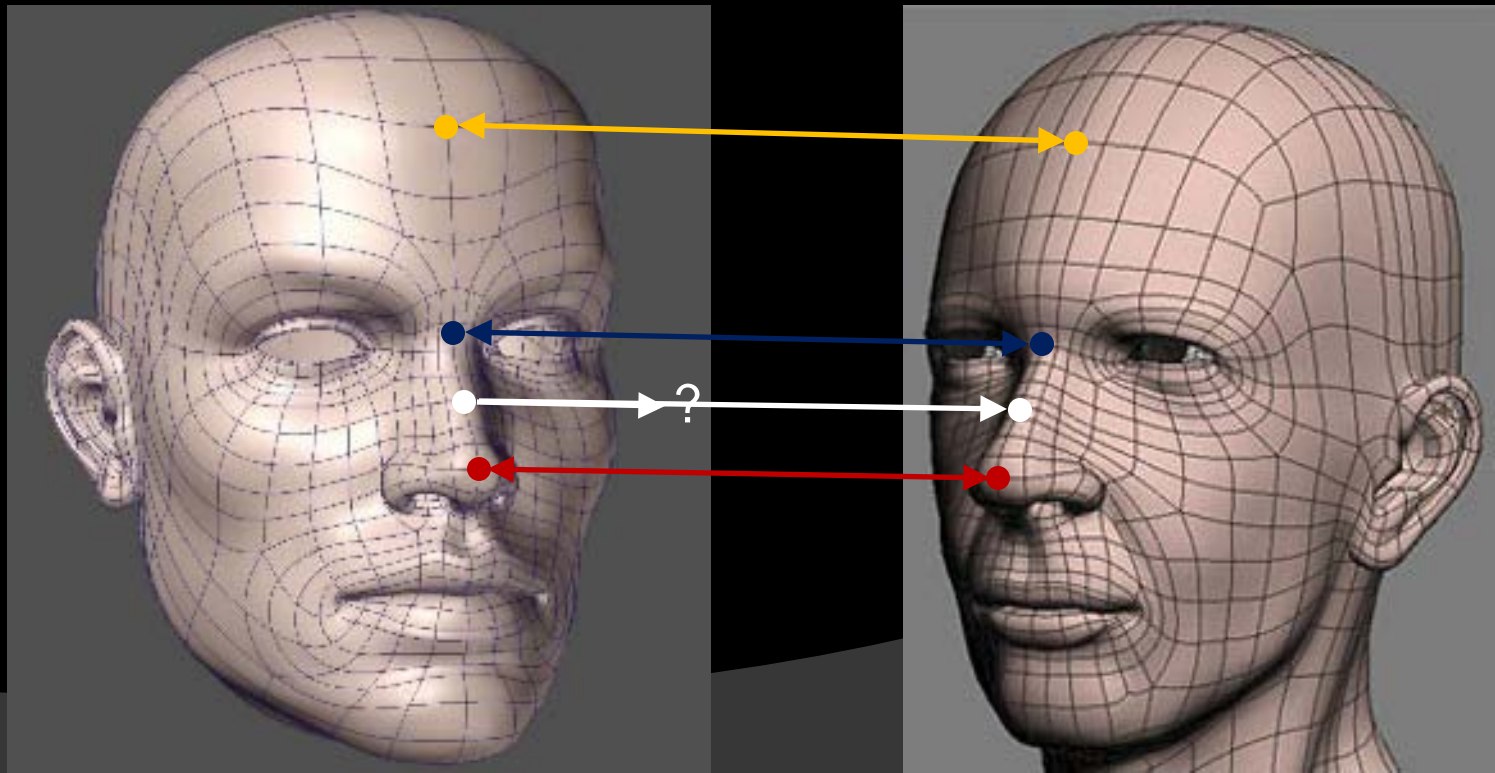


# Related Work

- ◎ **ICP** [Besl92, Feldmar96]
  - Global alignment, not landmark correspondence.
- ◎ **Mesh parameterisation** [Brett97,98; Lorenz99,00; Praun01, Davies02]
  - Re-mesh, rearrange mesh points consistently
  - Their landmark = re-parameterised mesh points  $\neq$  **facial** landmark.
- ◎ **Shape features** [Johnson99, Wang00, Yamany02, ]
  - Surface curvature, geodesic distance, spin image; not landmark correspondence.

# 3D Face Registration

- ◎ Main ideas of Hutton et al.:
  - Manually place 10 landmarks on training samples.
  - Use landmark correspondence to compute mapping.
  - Interpolate other points: **thin-plate spline**.

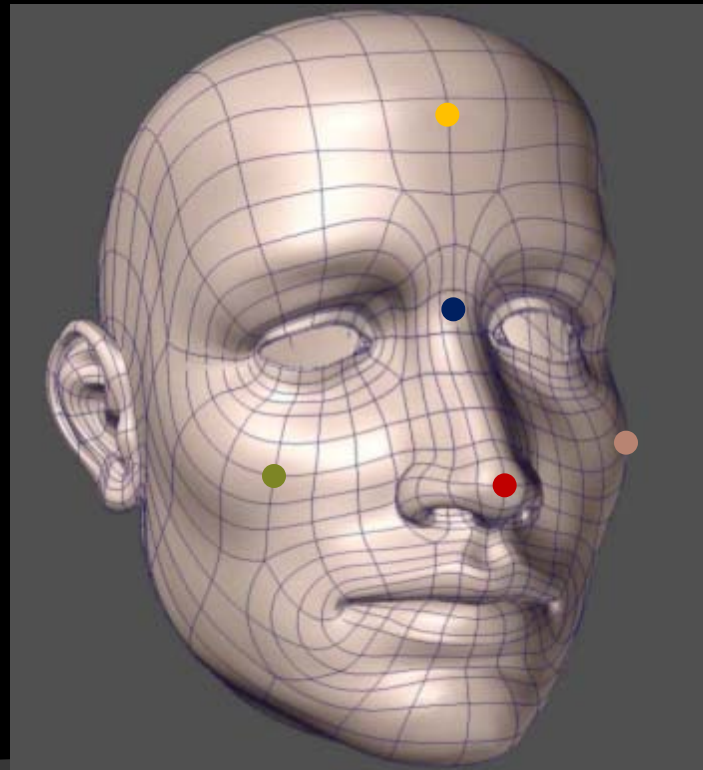


# Mean Landmarks

- ⦿ Compute mean landmarks of training samples.
- ⦿ Procrustes alignment:
  - Compute best alignment by **similarity transformation**, i.e., scaling, translation, rotation.
  - Align landmarks of all training samples.
  - Compute mean of landmarks.

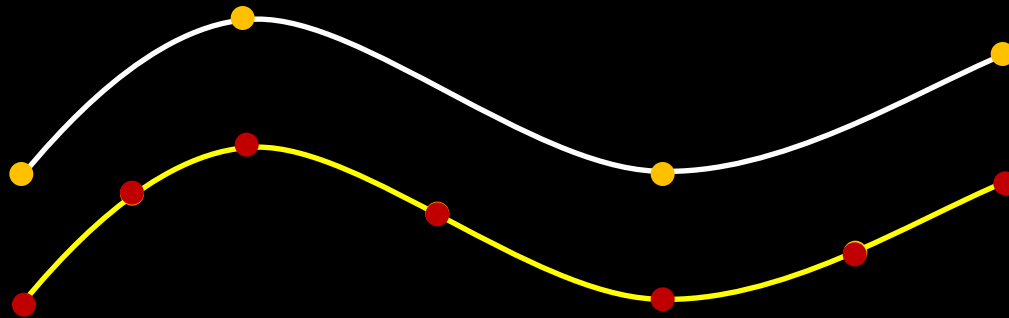
# Dense Correspondence

- ◎ Main steps:
  - Warp mesh by thin-plate spline so that landmarks coincide with mean landmarks.



# Dense Correspondence

- Resample warped mesh using reference mesh.
- Unwarp resampled mesh.
- Now, training samples have consistent mesh vertices.
- Some mesh vertices are facial landmarks.
- Now, can apply PCA on all mesh vertices.



# Statistical face model

## ◎ Main steps:

- Align all resampled training samples.
- Perform PCA.
- Keep top principal components.
- Normally,

$$\mathbf{x} = \bar{\mathbf{x}} + \Phi \mathbf{b}$$

shape parameters

- Hutton et al. used

$$\mathbf{x} = \bar{\mathbf{x}} + \Phi \mathbf{W} \mathbf{b}$$

unwhitening matrix

$$\mathbf{W} = \text{diag}(\sqrt{\lambda_1}, \dots, \sqrt{\lambda_k})$$

# Model Fitting

no facial landmark

- ◎ Fit mean shape  $\bar{x}$  to input shape  $y$ .
  - Apply ICP to align  $x$  to  $y$  (align global pose).
  - Repeat until convergence:
    - Map **vertices** on  $x$  to closest **surface points** on  $y$ .
    - New  $x_1$  has similar shape as  $y$ .
    - Align  $x_1$  to  $\bar{x}$  giving  $x_2$ .
    - Find shape parameters  $b$  of  $x_3$  wrt face model:

$$\mathbf{b} = \mathbf{W}^{-1} \Phi^T (\mathbf{x}_2 - \bar{\mathbf{x}})$$

- Restrict  $b$  to probable values  $b'$  according to model.
- Generate new shape  $x_3$  with  $b'$  from

close to  $y$

$$\mathbf{x}_3 = \bar{\mathbf{x}} + \Phi \mathbf{W} \mathbf{b}'$$

for generating  $y$

# Questions

- ⦿ Can it work for skulls?
- ⦿ How many skull landmarks?
- ⦿ Strengths?
- ⦿ Weaknesses?

