

3D face reconstruction from a single 2D image

Tan Chek Tien

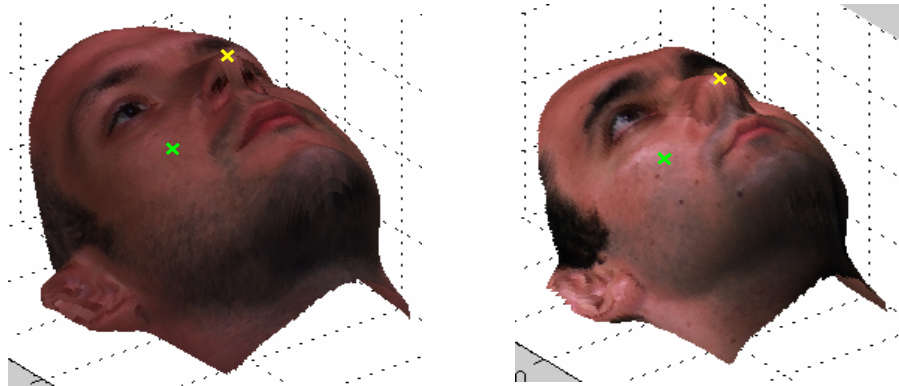
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- **Summary**

Morphable Model

– Problem Analysis

- Building the morphable model
 - Ill-posed problem



• Inputs

- Hi-res surface points
- Cheek, temples uniform
- Hard to match single points
- Corresponding points are
 - Near
 - Similar color
 - Similar surrounding shape

• Output

- Dense correspondence
- Shape/Texture

Morphable Model

– Problem Definition

Search for corresponding point in reference face to
example face model

- Assume all models aligned

- Inputs
 - r : 3D surface point in the reference face.
 - $\{p_i\}$: Set of 3D surface points in the example face model.
- Output
 - x : 3D surface point in $\{p_i\}$ that best matches r

Morphable Model

– Problem Definition

- **Problem Definition**

- Let $W(y)$ be a known function that defines a window of points centered at y .
- Let $S(y)$ be a known function providing shape information at y .
- Let $C(y)$ be a known function providing color information at y .

Hence, problem is to find the point x on the example face that minimizes E :

$$E = \lambda_C \sum_{r_i \in W(r), p_i \in W(x)} \|C(r_i) - C(p_i)\|^2 + \lambda_S \sum_{r_i \in W(r), p_i \in W(x)} \|S(r_i) - S(p_i)\|^2$$

where $\lambda_p, \lambda_C, \lambda_S$ are weighting factors

And $x \in R(r)$ where $R(r)$ defines a region around r

Morphable Model

– Algorithm

- Search Process

Reference



Example



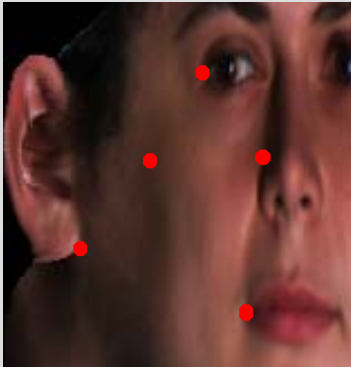
- $C(y)$ – RGB values
- $S(y)$ – Gaussian Curvature

Morphable Model

– Algorithm

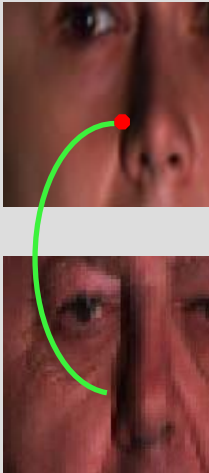
- Enhancement

Feature Detection



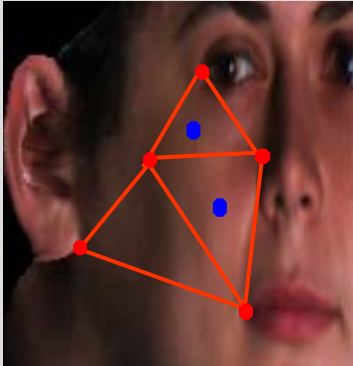
- Corner detection
- By Gaussian Curvature

Matching



- Min E

Interpolation

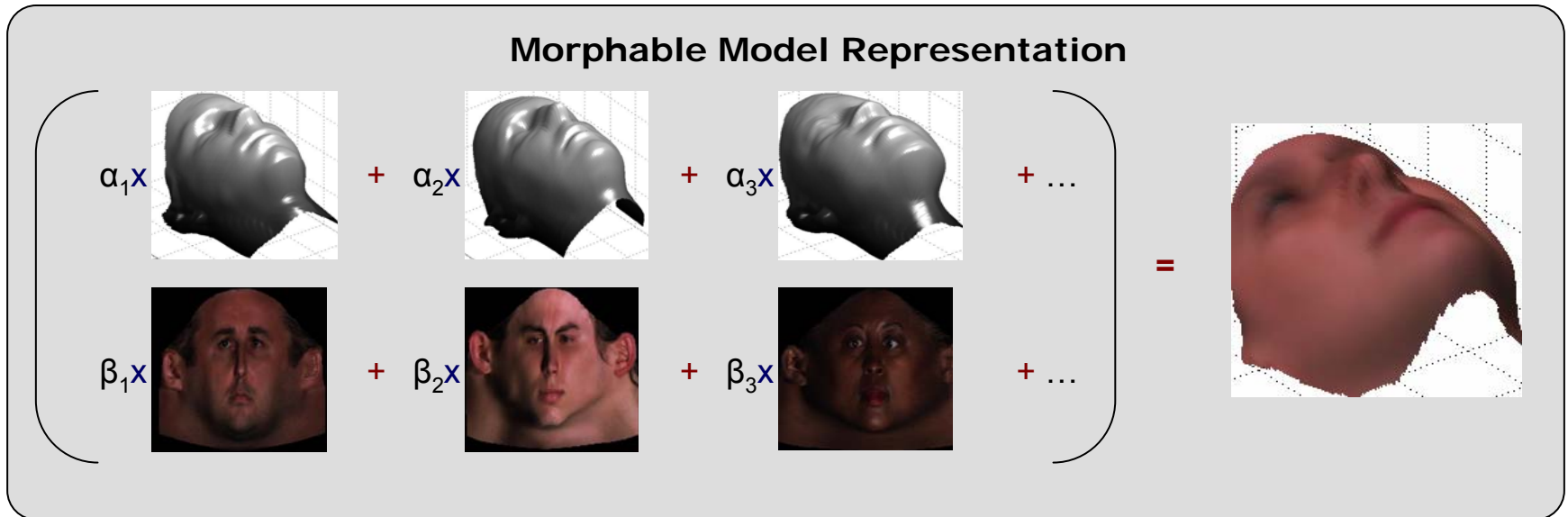


- Delaunay Δ
- Interpolation

Morphable Model

– Algorithm

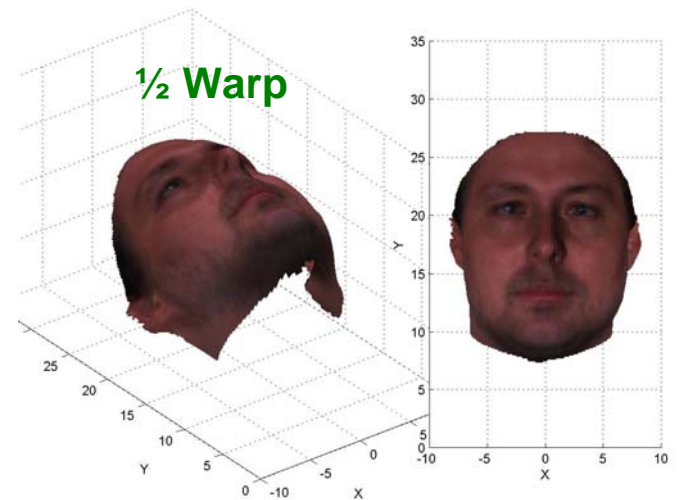
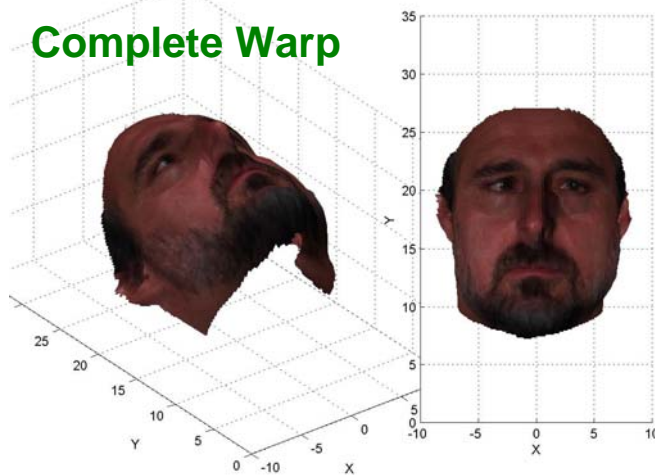
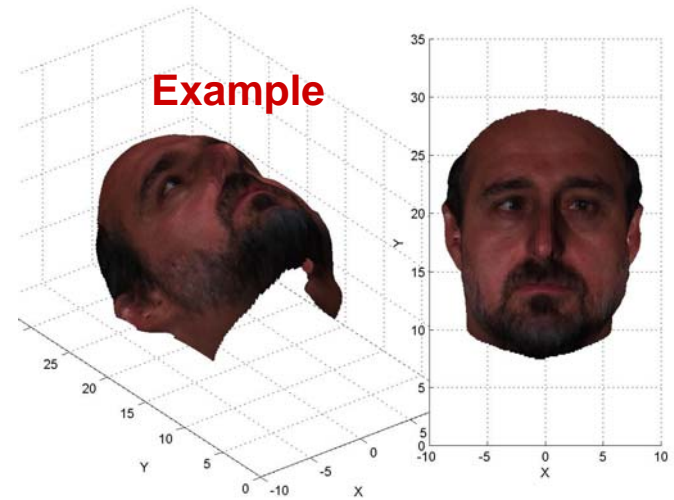
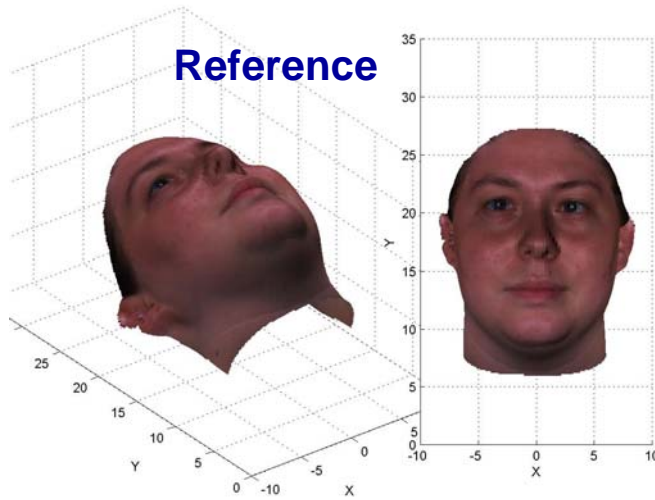
- Deformable Model



Morphable Model

– Results

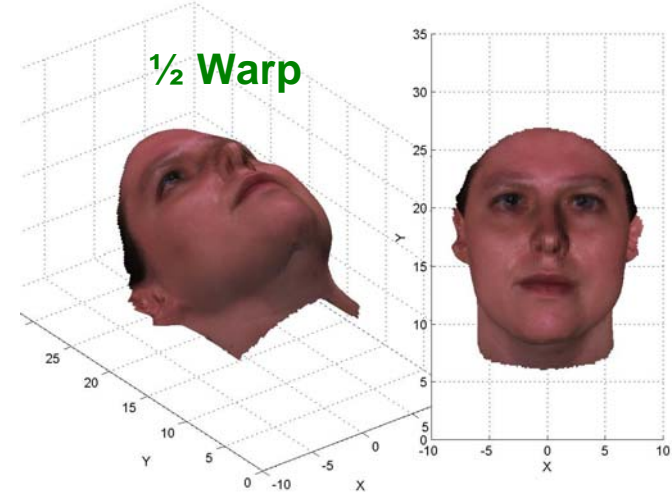
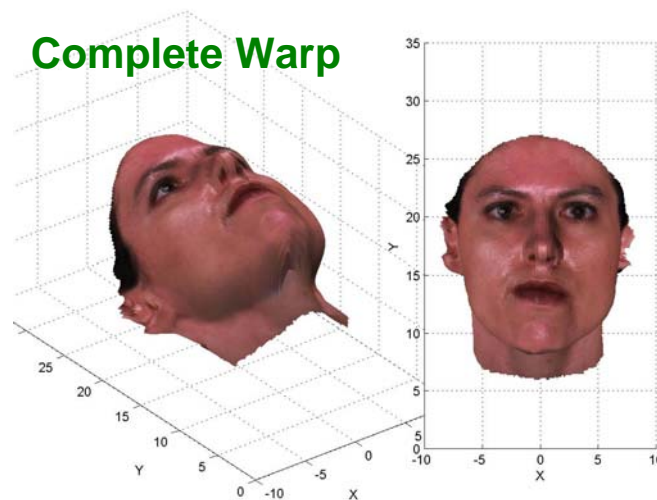
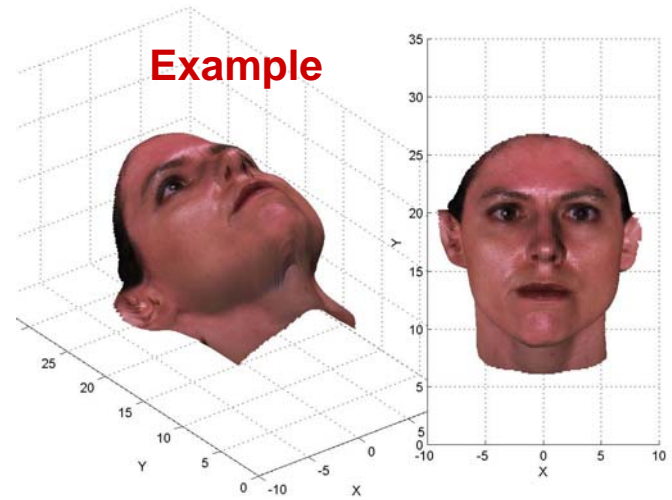
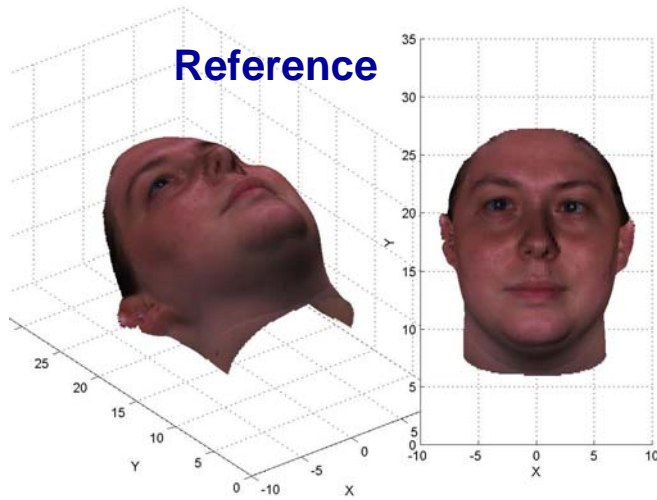
- +ve



Morphable Model

- Results

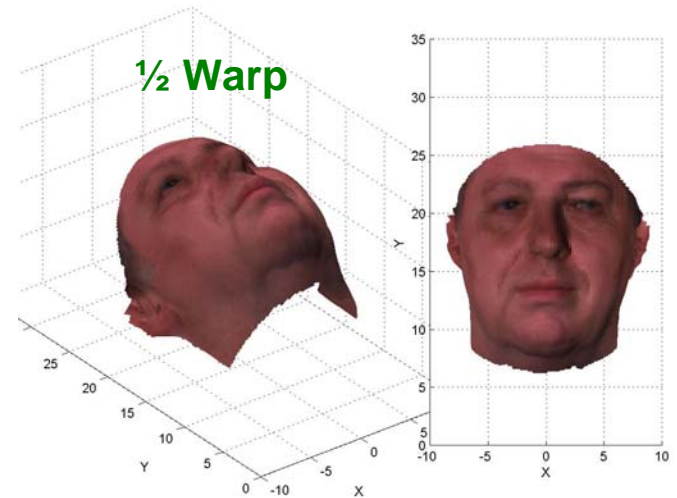
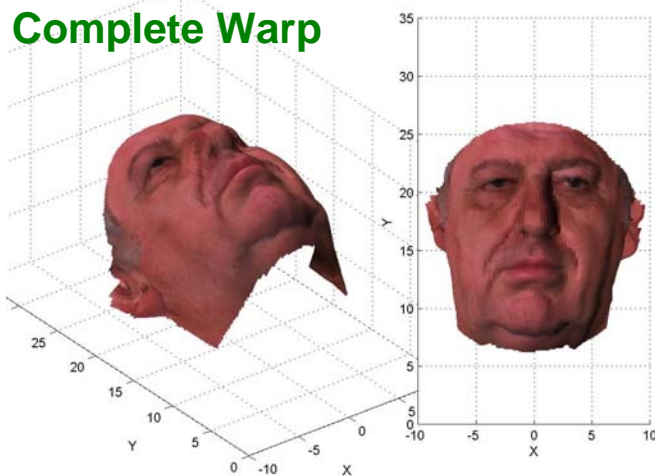
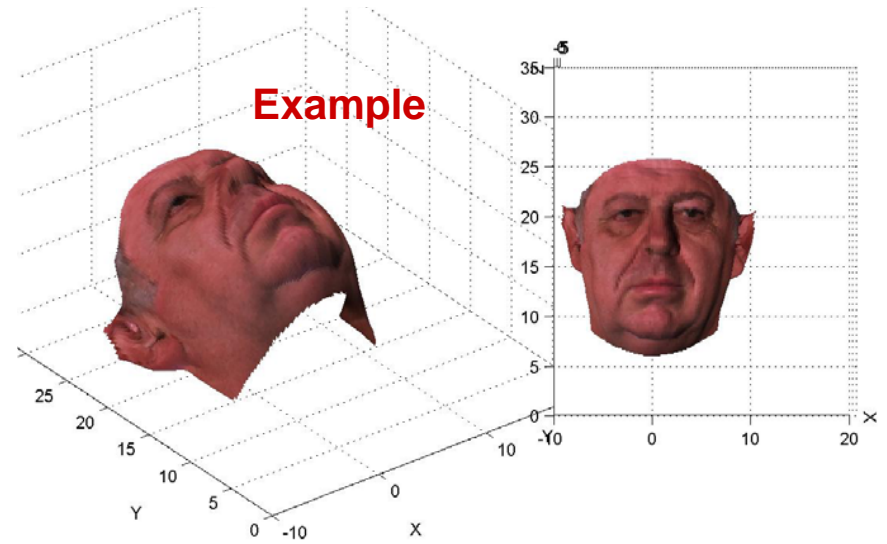
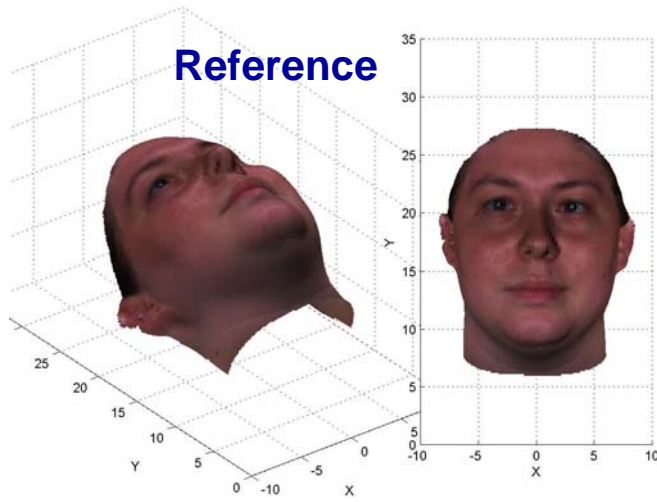
- +ve



Morphable Model

- Results

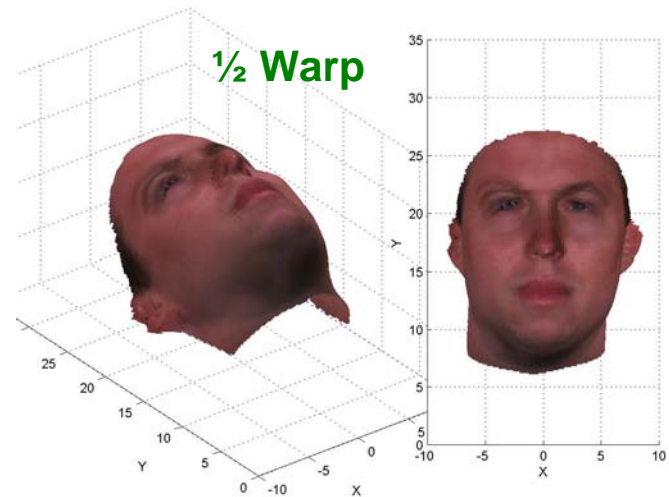
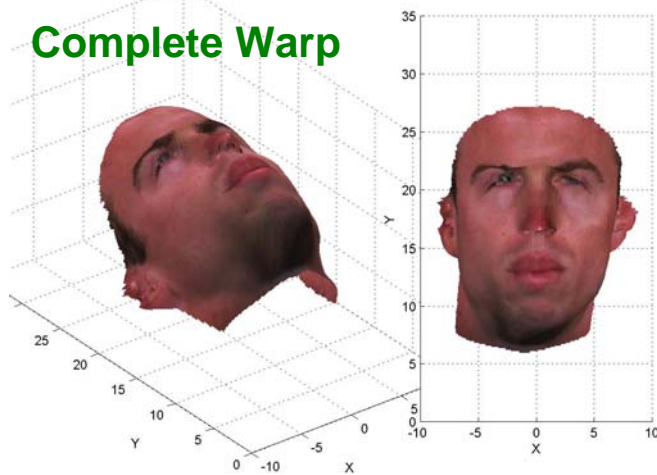
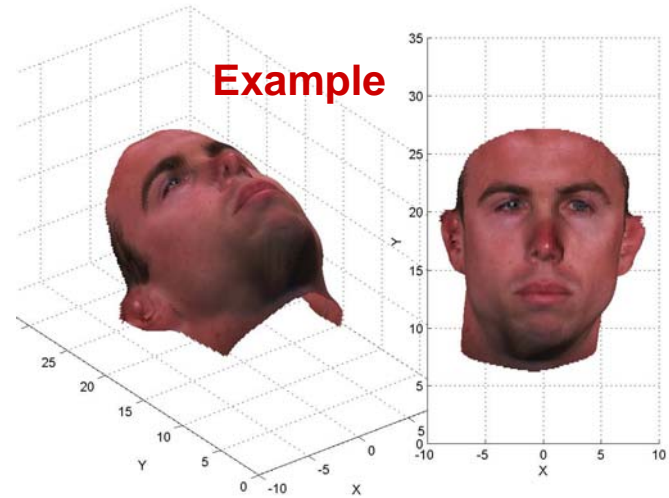
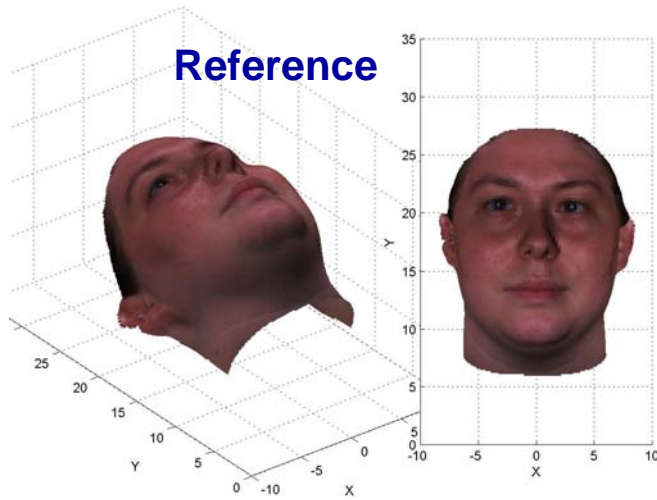
- +ve



Morphable Model

- Results

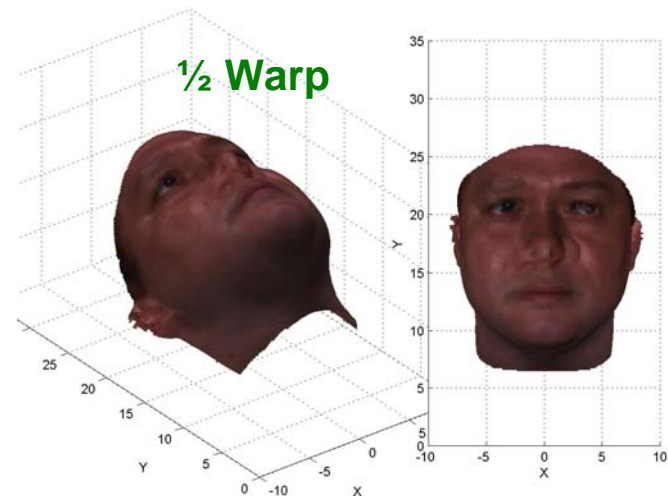
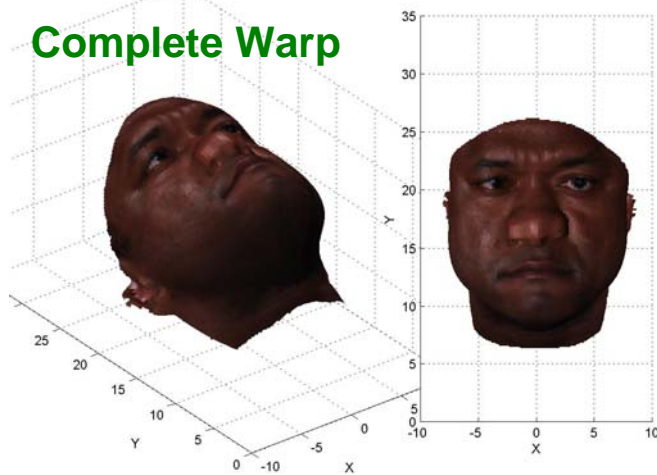
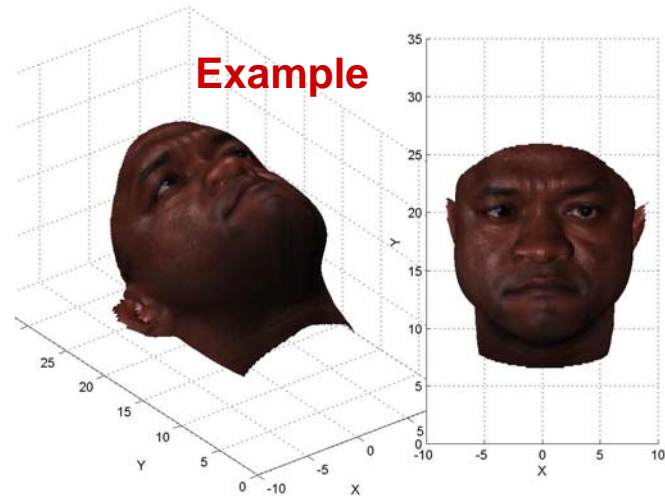
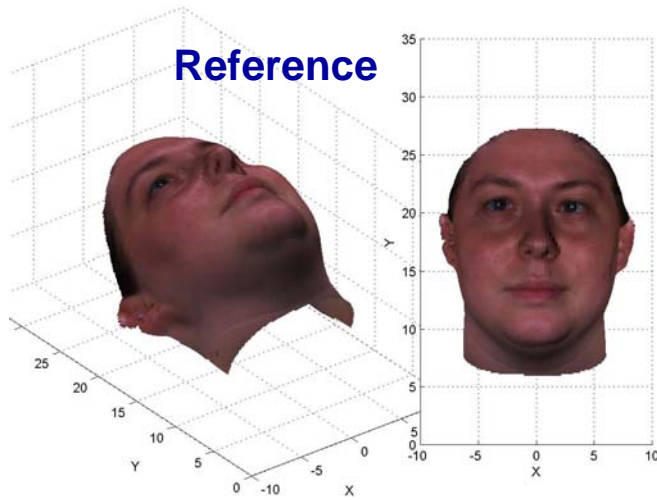
- ve



Morphable Model

- Results

- ve



Model Fitting

– Problem Analysis

Input



Single 2D facial image.

- Contains unwanted regions
- Illumination, Pose, Camera
- Assume no facial artifacts



Deformable Model

- Interact with surroundings
- Within Face Space

Specify your own

Synthesize

S1	<	>	-112.2911	T1	<	>	0.2226e
S2	<	>	-56.4926	T2	<	>	0.0
S3	<	>	-43.9871	T3	<	>	0.0
S4	<	>	0.0	T4	<	>	0.0
S5	<	>	0.0	T5	<	>	0.0
S6	<	>	0.0	T6	<	>	0.0
S7	<	>	0.0	T7	<	>	0.0
S8	<	>	0.0	T8	<	>	0.0

Shape Components Texture Components



Output



3D face model of person in image(s)

- Should look like the person
- Limited angle of view



Model Fitting

– Problem Definition

Registration of a deformable model to an image with known correspondences

- **Inputs**

- Set of 2D points, $\{e_j\}$, on the 2D input image.
- Set of 3D points, $\{m_i\}$, on the deformable model

- **Outputs**

- $\{m'_i\}$, the deformed version of $\{m_i\}$

Model Fitting

– Problem Definition

- **Problem Definition**

- Let f be a known correspondence function from $\{m_i\}$ to $\{e_j\}$,
i.e. $f(m_i) \in \{e_j\}$
- Let D be a deformation function such that optimally $m_i' = D(m_i)$
- Let T be a rigid transformation function on the model
- Let P be a known projection function on the model
- Let R be a rendering function on the model which includes camera intrinsic params and illumination params
- Let S be a known function providing shape information.

Hence, problem is to find D, T and R that minimizes E :

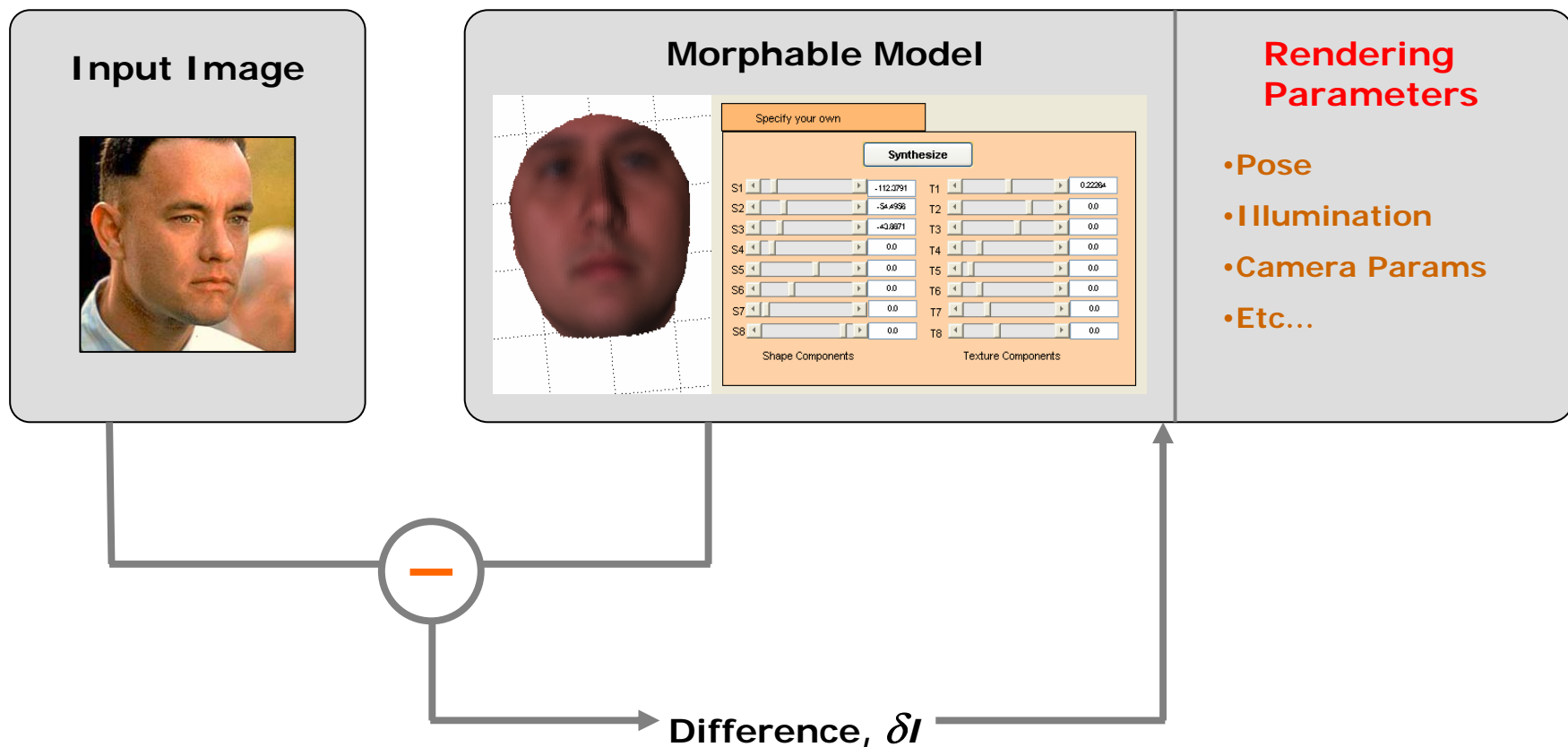
$$E = \lambda \sum_i \|P(R(T(D(m_i)))) - f(m_i)\|^2 + \cancel{(1-\lambda) \sum_i \|S(m_i) - S(D(m_i))\|^2}$$

where λ is a weighting factor

Model Fitting

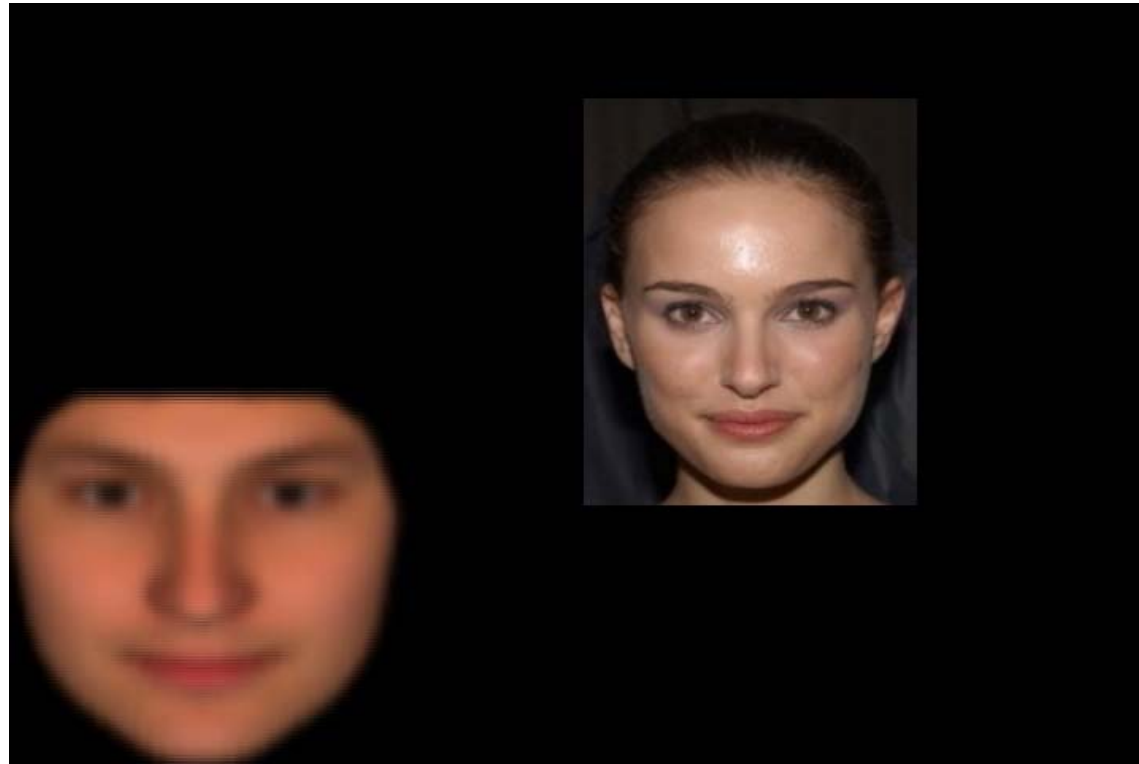
– Algorithm

- Optimization Process
 - Stochastic Newton Optimization [2] [5]



Model Fitting

– Results

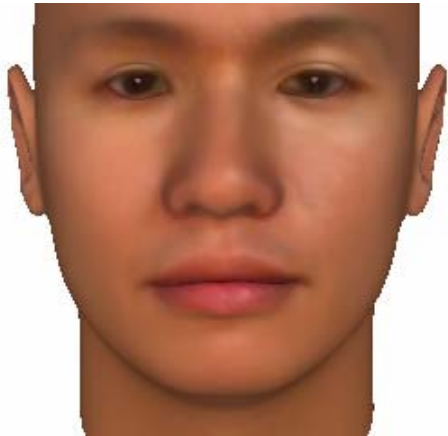


Model Fitting

– Results

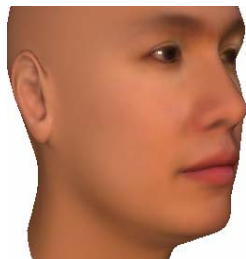


Model Fitting – Results



Summary

- Accuracy dependant on how well variance is captured in 3DMM
 - Database must be large and well distributed.
- Comparison with Blanz and Vetter [1]



This	Blanz & Vetter [1]
3DMM has curvature	3DMM only by position and texture
3DMM – 90 faces	3DMM – 200 faces
~10mins	~4.5mins
Texture dependant on 3DMM	Texture extracted from image

References

- [1] A Morphable Model for the Synthesis of 3D Faces (1999)
 - V.blanz and T. Vetter
- [2] Face recognition based on fitting a 3D morphable model (2003)
 - V.blanz and T. Vetter
- [3] Equivalence and Efficiency of image alignment algorithms (2001)
 - S. Baker and I. Matthews
- [4] Three-Dimensional Correspondence (1998)
 - C. Shelton
- [5] FaceGen (1998)
 - Singular Inversions
- [6] Metrics and Visualization Tools for Surface Mesh Comparison. (2001)
 - Zhou, L and Pang, A.