

# A morphable model for the synthesis of 3D faces

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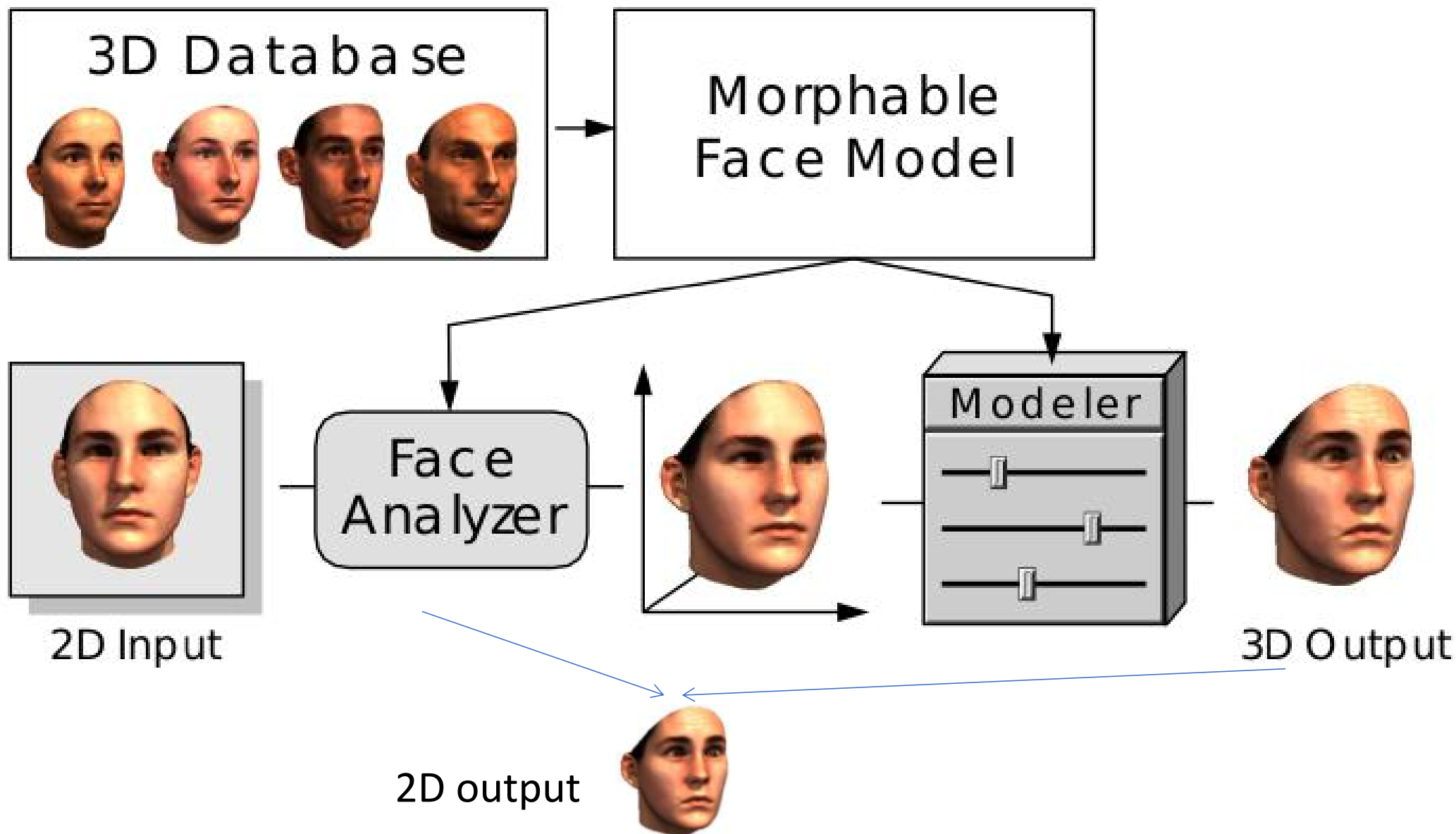
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The slides and a list of references can be found from  
<https://github.com/fwcore/object-detection>

# Outlines

- **2D-to-3D and 3D-to-2D workflow**
- **3D morphable face model**
  - build model from 3D face database
  - modeling facial attributes
  - 3D correspondence using optic flow
- **Matching a morphable model to images**
  - matching 3D scan
  - matching 2D image

# 2D-to-3D and 3D-to-2D workflow

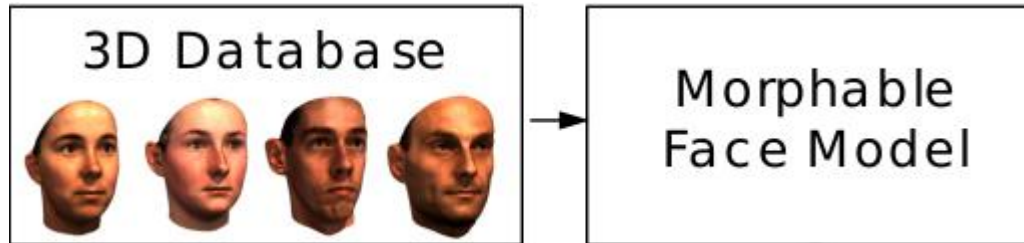


Blanz & Vetter (1999) A morphable model for the synthesis of 3D faces.

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# 3D morphable face model



## 3D database (1999)

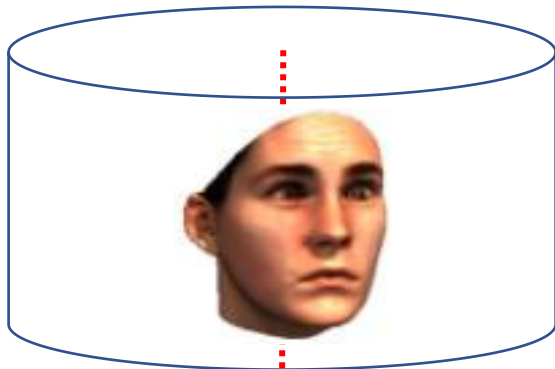
Cyberware™ laser scans

100 young male and 100 young female

$r(h, \phi)$ ,  $R(h, \phi)$ ,  $G(h, \phi)$ ,  $B(h, \phi)$

512 equally-spaced in  $\phi$ , ~70,000 vertices

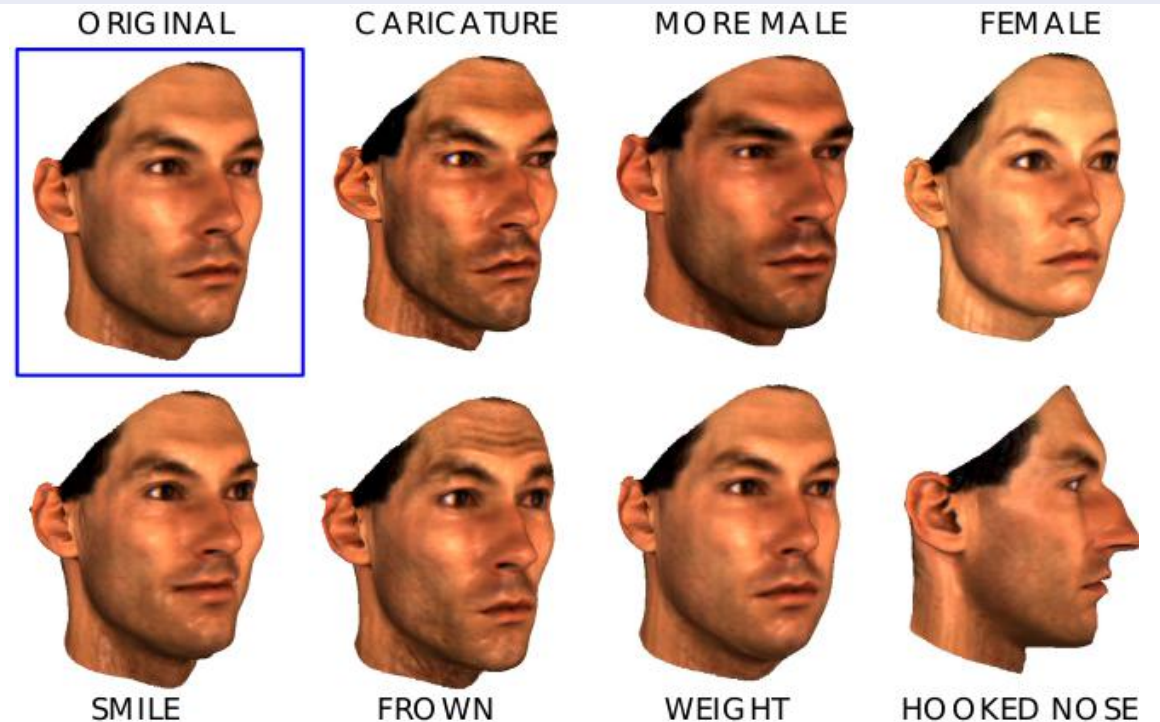
no makeup, accessories, facial hair. Wearing bathing cap, and removed digitally.



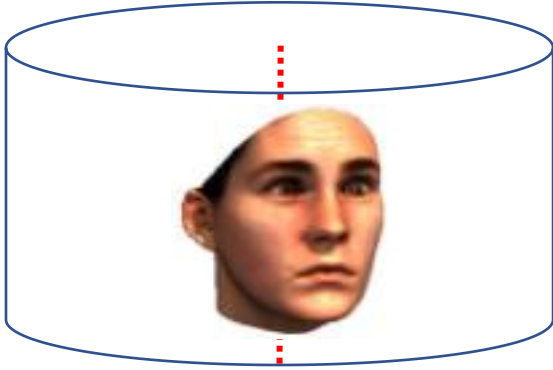
## Spec. of morphable face model (1999)

being real face-like

easy to change facial attributes



# build model from 3D face database



## 3D face

- shape vector ( $x_1, y_1, z_1, x_2, y_2, z_2, \dots$ )
- texture vector ( $R_1, G_1, B_1, R_2, G_2, B_2, \dots$ )

## 3n+3n dimensional vector

n: number of vertices

m: number of faces in database

$m \ll n$

## Assumptions:

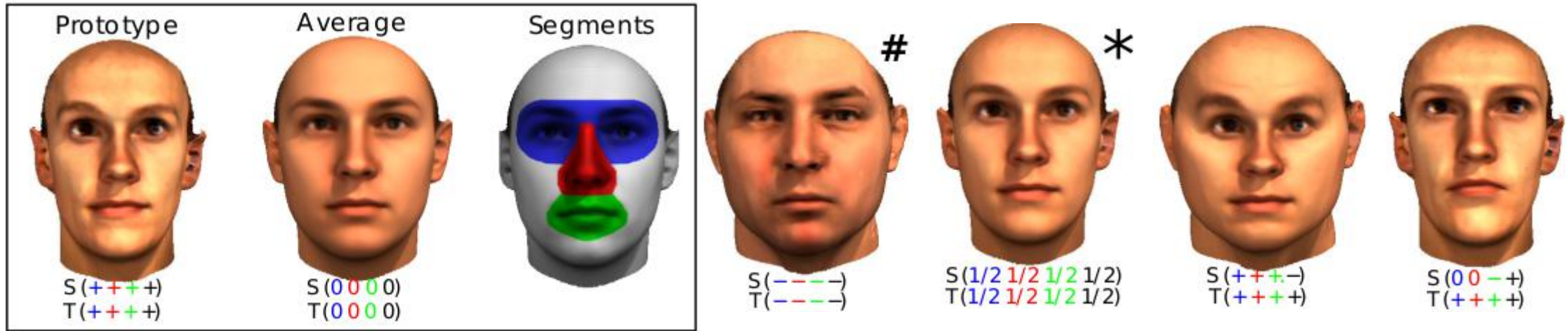
- Gaussian distribution in 6n dim. space
- shape and texture are independent (to generate more faces)
- *face subregions are independent (to further increase the possible faces). Require blending the borders later on.*

## Gaussian = PCA:

PCs are decorrelated and Gaussian distributed.

$$S_{model} = \bar{S} + \sum_{i=1}^{m-1} \alpha_i s_i, \quad T_{model} = \bar{T} + \sum_{i=1}^{m-1} \beta_i t_i$$
$$p(\vec{\alpha}) \sim \exp\left[-\frac{1}{2} \sum_{i=1}^{m-1} (\alpha_i / \sigma_i)^2\right],$$

# modeling facial attributes: linear model



$$\Delta S = \sum_{i=1}^m \mu_i (S_i - \bar{S}), \quad \Delta T = \sum_{i=1}^m \mu_i (T_i - \bar{T})$$

$\mu_i$ : facial attribute (obtained by regression)

gender, fullness of faces, darkness of eyebrows,  
double chins, hooked vs. concave noses.

# Preparing database: 3D correspondence using optic flow

**Key problem:** build a point-to-point correspondence between vertices of faces.

**Mathematically:** consistently label vertices by 1 to  $n$

## 3D scans

- $r(h, \phi)$
- $R(h, \phi), G(h, \phi), B(h, \phi)$

## 3D face

- shape vector  $(x_1, y_1, z_1, x_2, y_2, z_2, \dots)$
- texture vector  $(R_1, G_1, B_1, R_2, G_2, B_2, \dots)$

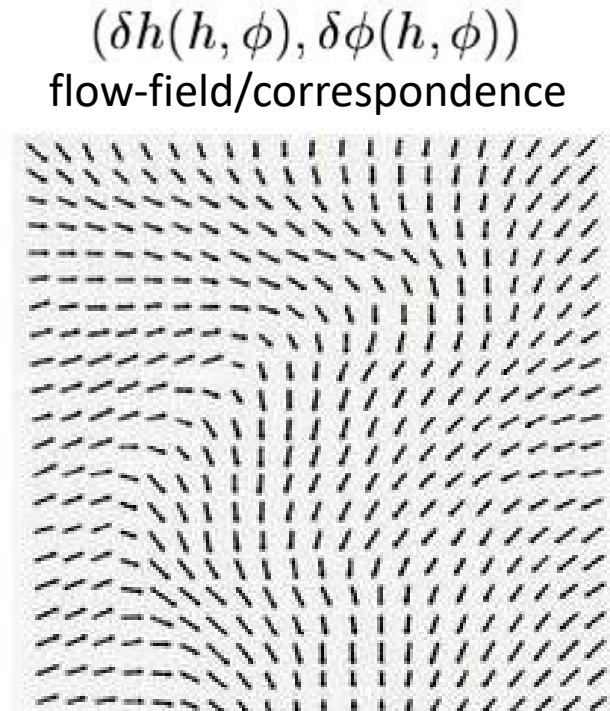
?



**minimize**

$$\|\mathbf{I}_1(h, \phi) - \mathbf{I}_2(h + \delta h, \phi + \delta \phi)\|$$

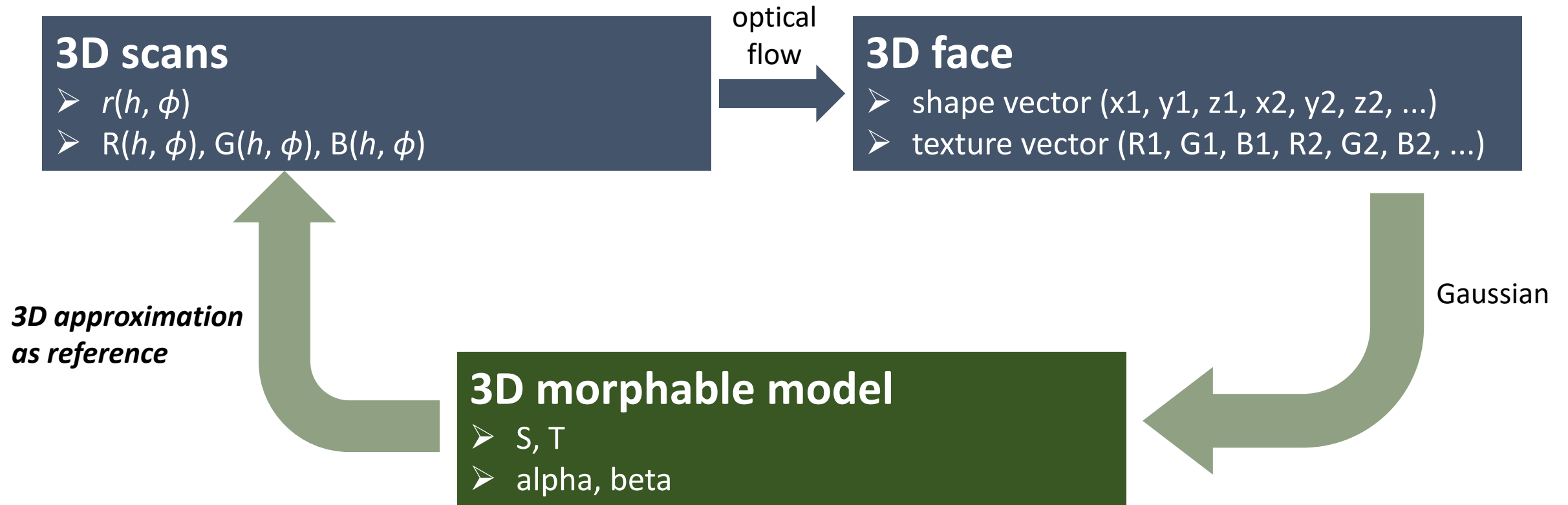
+  
quadratic coupling  
between  
neighboring flow



$$\mathbf{I}(h, \phi) = (R(h, \phi), G(h, \phi), B(h, \phi), r(h, \phi))^T + \text{mean curvature (optional)}$$



# Preparing database: bootstrapping



We need tech to approximate 3D scans by 3D morphable model.

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# matching 3D scan

mapping  $\tilde{C} : \mathcal{R}^3 \rightarrow \mathcal{R}^2, \quad (x, y, z) \mapsto (h, \phi)$

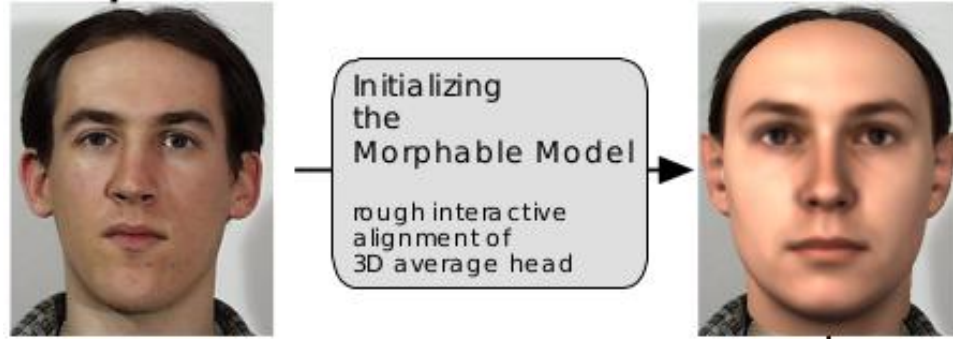
$$\mathbf{I}(h, \phi) = (R(h, \phi), G(h, \phi), B(h, \phi), r(h, \phi))^T$$

*minimize by varying  $\alpha, \beta$*

$$E = \sum_{h, \phi} \|\mathbf{I}_{input}(h, \phi) - \mathbf{I}_{model}(h, \phi)\|^2$$

# matching 2D image

## 2D Input



Automated 3D Shape and Texture Reconstruction



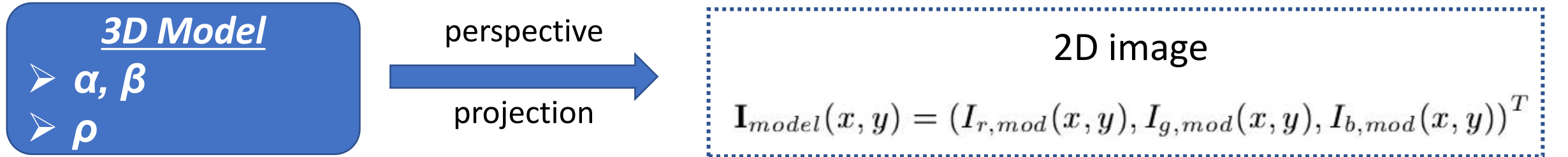
Illumination Corrected Texture Extraction



## 3D Model parameters

- $\alpha, \beta$
- **rendering  $\rho$ :**
  - camera position,
  - object scale,
  - image plane rotation & translation,
  - intensity of ambient light,
  - intensity of directed light,
  - color contrast, offset, and gain in RGB channel
  - (fixed & estimated by user) camera distance
  - (fixed & estimated by user) light direction
  - (fixed & estimated by user) surface shininess

# matching 2D image

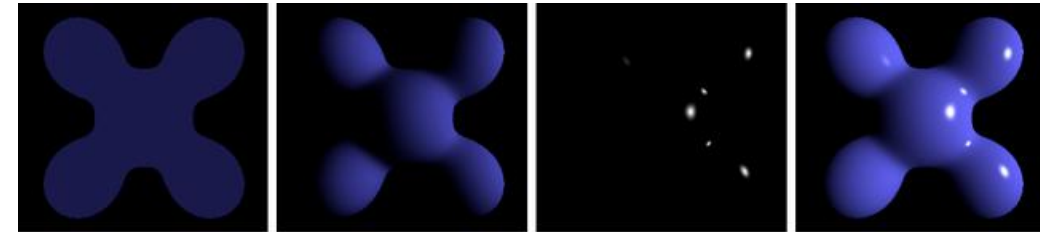


$$I_{r,model,k} = (\underbrace{i_{r,amb}}_{\text{ambient light}} + \underbrace{i_{r,dir} \cdot (\mathbf{n}_k \mathbf{l})}_{\text{directed light}}) \bar{R}_k + \underbrace{i_{r,dir} s \cdot (\mathbf{r}_k \mathbf{v}_k)^\nu}_{\text{specular reflection}}$$

ambient  
light

directed  
light

specular  
reflection



Ambient + Diffuse + Specular = Phong Reflection

$$E_I = \sum_{x,y} \|\mathbf{I}_{input}(x, y) - \mathbf{I}_{model}(x, y)\|^2$$

**minimize**  $E = \frac{1}{\sigma_N^2} E_I + \sum_{j=1}^{m-1} \frac{\alpha_j^2}{\sigma_{S,j}^2} + \sum_{j=1}^{m-1} \frac{\beta_j^2}{\sigma_{T,j}^2} + \sum_j \frac{(\rho_j - \bar{\rho}_j)^2}{\sigma_{\rho,j}^2}$

L2 regularization

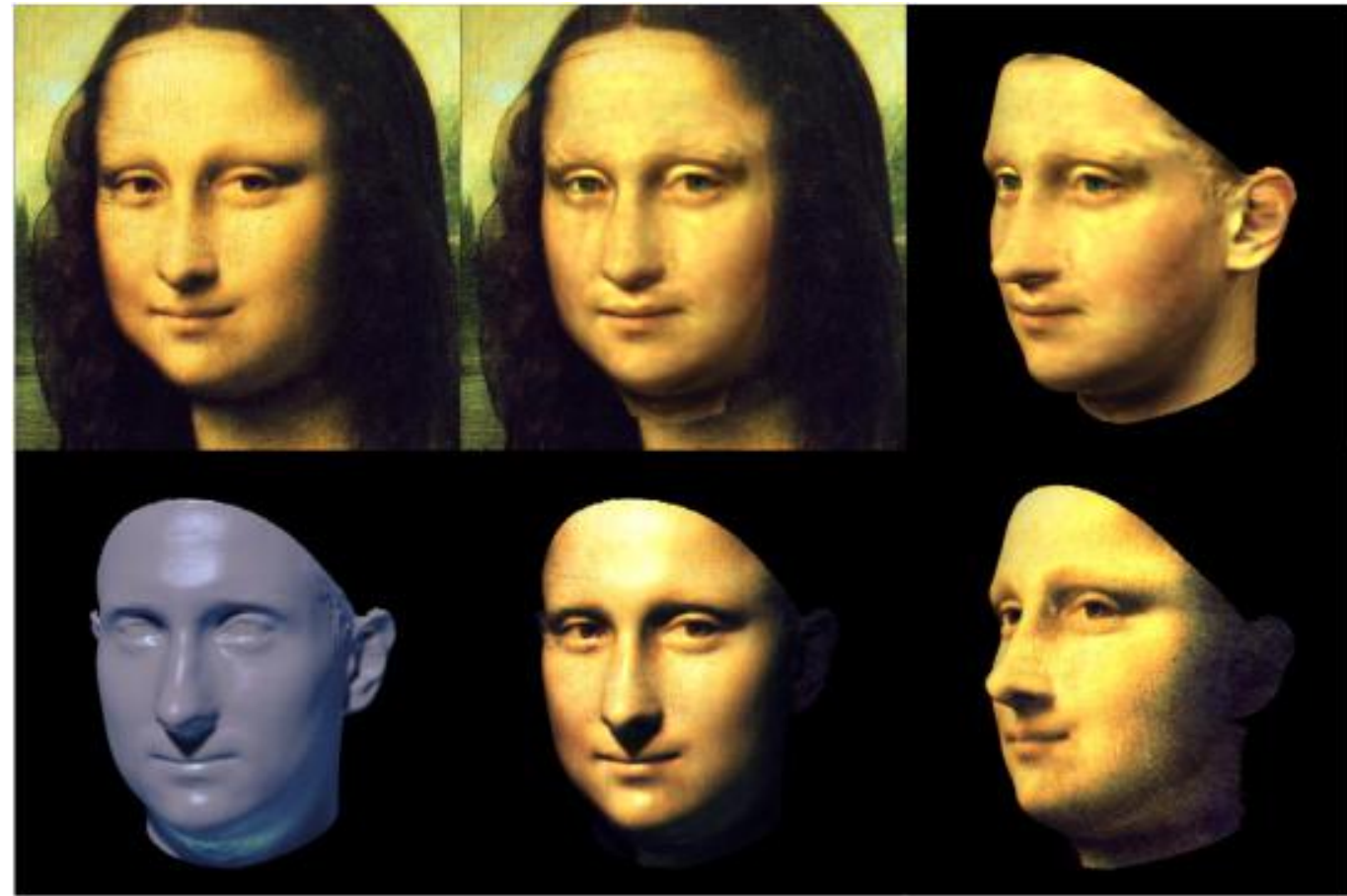
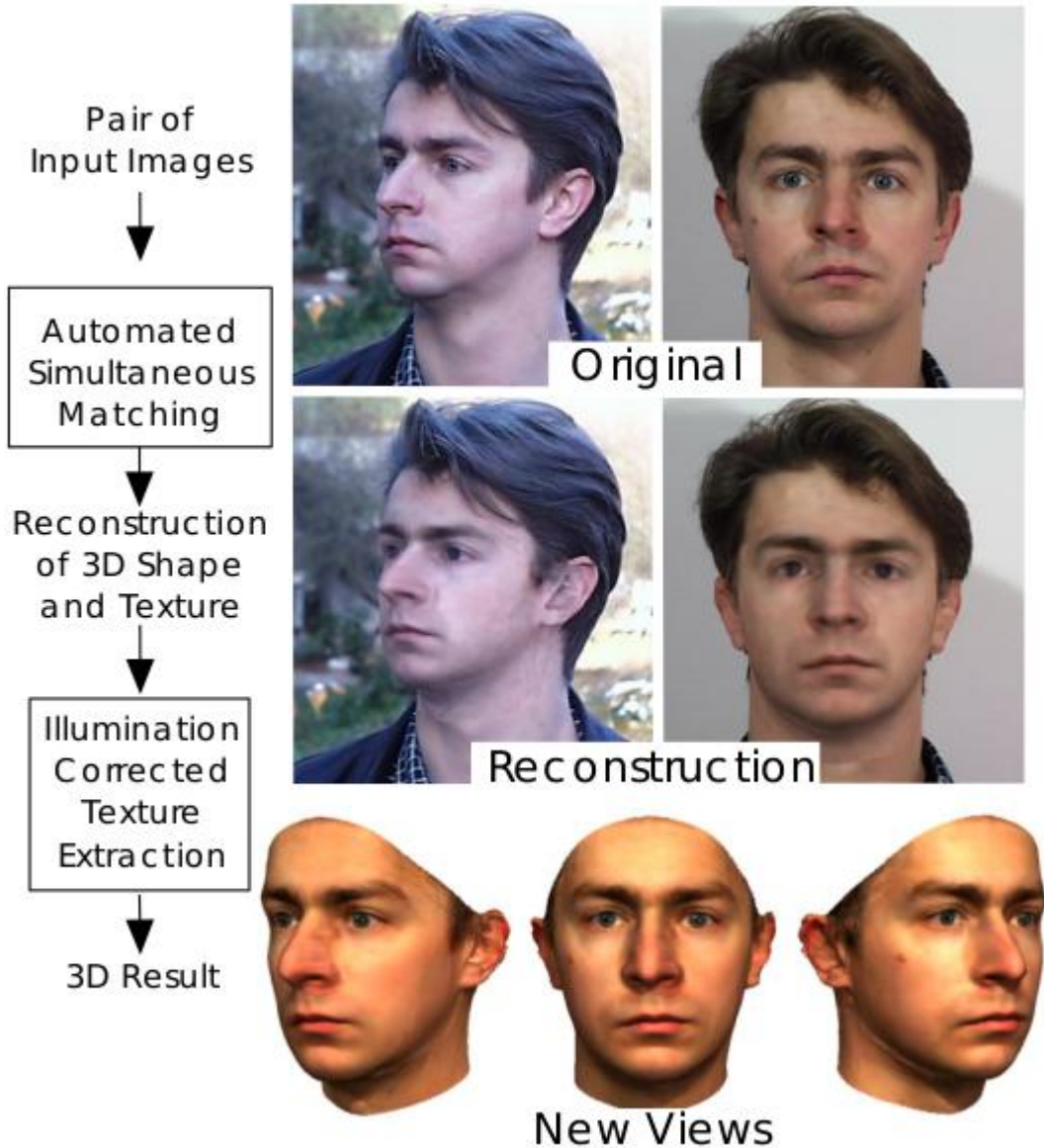
# Coarse-to-fine

1. start iteration with down-sampled input image and low-resolution 3D model.
2. slowly add more  $\alpha$ ,  $\beta$  and principle components during optimization
3. start with strong regularization and then slowly reduce it
4. finally, fixed  $\rho$  and change to segmented 3D model to improve details
5. illumination-corrected texture extraction: extract raw-image details and overlay.



# Examples

two -> one



Performance: 50 min on SGI R1000 processor (1999)



Original



Initialization



3D Rec onstruc tion



Re construction  
of Shape & Texture



Texture Extraction  
& Fa cial Expression



Cast Shadow



New Illumination



Rotation



# Open questions

## 1. Extending the database:

- mainly Caucasian of middle age
- facial expressions, visemes, face variation during speech (limited by the slow laser scan  $\sim 10$  sec/scan)

## 2. Extending the model:

- include hair