

AGENCE NATIONALE DE LA RECHERCHI

3D FACE ANALYSIS AND RECOGNITION

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Why 3D shape of face?

2D technology limitations

Pose and scale variationsIllumination variations

• New 3D Challenges

- Facial deformations
- Data quality
 - Holes/noise/resolution



Scientific challenge

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• *Elastic deformations* (stretching/shrinking)





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Introduction and motivation

13D face recognition using radial curves

- Method overview
- 3D face shape analysis: elastic metric
- o Evaluation experiments
- Hierarchical organization

2 Partial biometric

- o **3D nose analysis**
- o Experimental results
- Hierarchical organization

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Facial Curve Parameterization

Why q?

• $q \in IL^2$ (Hilbert space)

The IL^2 norm of q is the length of the curve C.

□ In order to let the curve stretch, shrink and bend freely









Geodesic distance computation

between curves

Input: Two curves β^1 and β^2

1- For each i = 1, 2 compute the Square Root velocity Function $q^{i}(t) \doteq \frac{\dot{\beta}^{i}(t)}{\sqrt{\|\dot{\beta}^{i}(t)\|}}$

2- Compute the optimal rotation and re-parametrization O^* and γ^*

$$(O^*, \gamma^*) = \arg .min_{(O,\gamma) \in SO(3) \times \Gamma} d_c(q^1, \sqrt{\dot{\gamma}} O(q^2 \circ \gamma))$$

3- $q_2^*(t) = \sqrt{\dot{\gamma}^*(t)} O^* q^2(\gamma^*(t)))$ **4-** Compute the geodesic distance between q^1 and q_2^* . $d_c(q^1, q_2^*) = \cos^{-1}(\langle q^1, q_2^* \rangle)$ **5-** $d_s([q^1], [q^2]) = d_c(q^1, q_2^*)$ **Output:** $d(\beta^1, \beta^2) = d_s([q^1], [q^2])$











The distance between S_1 and S_2 is the sum of distances between curves from the same level :

$$d_{S}: C^{[0,L]} \times C^{[0,L]} \to \mathfrak{R}^{+}$$
$$d_{S}(S_{1}, S_{2}) = \int_{0}^{L} d(c_{\lambda}^{1}, c_{\lambda}^{2}) d\lambda$$

Distance between two facial surfaces





Robustness to Missing data/Pose

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- Radial curves extraction
- Criterion for curve quality inspection

 (i) # of connected components = 1
 (ii) Curve length > T (threshold)

Curves quality inspection



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Evaluation experiments

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• GavabDB*

- Public <u>http://www.gavab.etsii.urjc.es/recursos_en.html</u>
- The most expression-rich and noise-prone 3D face dataset
- 61 subjects (45 male + female) Caucasian
- 9 sessions/subject
 - 2 x Neutral frontal sessions
 - 3 x Expressive sessions
 - 4 x Neutral non-frontal sessions

Expression variation

Pose variation



*A. B. Moreno and A. Sánchez. Gavabdb: a 3d face database. In Workshop on Biometrics on the Internet, pages 77–85, Vigo, March 2004.

Evaluation experiments

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• *Identification results and comparison with state-of-the-art approaches*

	Li et al. [8]	Moreno et al. [10]	Mahoor et al. [9]	Berretti et al. [1]	Mousavi et al. [11]	Our
(a)	96.67%	90.16%	95%	94%	-	100%
(b)	93.33%	77.9%	72%	81%	-	94.54%
(c)	94.68%	-	78%	84.25%	91%	94.67%
(d)	-	-	85.3%	80%	-	100%
(e)	-	-	88.6%	79%	-	98.36%
(f)	-	-	-	82.66%	81.67%	96.99%
(g)	-	-	-	-	-	70.49%
(h)	-	-	-	-	-	86.89%

(a) Neutral, (b) Expressive, (c) Neutral + Expressive, (d) Rotated looking down, (e) Rotated looking up, (f) Overall, (g) Right profile, (h) Left profile

• Examples of correct and incorrect recognition



BMVC 2010



Evaluation experiments

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Identification results and comparison with state-of-the-art approaches

Recognition results comparison of different methods on FRGCv2 dataset (neutral vs. all senario).

Approaches	Berretti et al. [4]	Chang et al. [10]	Mian et al. [26]	Faltemier et al. [14]	Our approach
Rank-1	94.1%	94.6%	96.2%	98.1%	97.7%

[14] Faltemier et al. IEEE Transactions on Information Forensics and Security 2008
[4] Berretti et al. IEEE Transactions on Pattern Analysis and Machine Intelligence 2010
[10] Chang et al. IEEE Transactions on Pattern Analysis and Machine Intelligence 2006
[26] Mian et al. IEEE Transactions on Pattern Analysis and Machine Intelligence 2007

Submitted to IEEE Transaction PAMI 2010



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Open platform for 3D Face Recognition

- Face Analysis and Recognition using 3D (FAR3D) financed by the French National Research Agency (ANR)
 - http://www-rech.telecom-lille1.eu/far3d/
 - × 2008-2010
 - University Lille1 (TELECOM Lille1) coordinator
 - Ecole Centrale de Lyon
 - Eurecom Institute
 - Thales R&T



3D Body scanning Lugano 2010

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Conclusions and future directions

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Scientific contributions

- Propose original 3D face recognition algorithms robust to expressions and missing data (pose and data quality)
- Mathematical tools for statistics (mean facial shape, etc.)
 - × Hierarchical organization of gallery
- Develop *an open technology platform* that integrates matching algorithms based on 3D facial shape
- Future directions/Open problems
 - **On-the-fly** 3D face recognition algorithms.
 - Facial attributes recognition using 3D (age estimation, expression, ethnicity, gender, etc.)



Thank you for your attention Questions



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Publications related to this work

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1-H. Drira, B. B. Amor, A. Srivastava, and M. Daoudi. A Riemannian analysis of 3d nose shapes for partial human biometrics. In *IEEE International Conference on Computer Vision, 2009.*

2-H. Drira, B. B. Amor, M. Daoudi, and A. Srivastava. Nasal Region Contribution in 3D Face Biometrics Using Shape Analysis Framework In IEEE/IAPR International Conference on Biometric, 2009.

3- H. Drira, B. Benamor, M. Daoudi, and A. Srivastava. Pose and expression-invariant 3d face recognition using elastic radial curves. In *Proceedings of the* **British Machine** *Vision Conference*, 2010.

4-H. Drira, B. Ben Amor, M. Daoudi, A. Srivastava, Elastic Radial curves to model 3D face deformations , in 3DOR workshop, ACMMMultimedia 2010 (ACMMM) 2010.

4- H. Drira, B. Benamor, M. Daoudi, and A. Srivastava. Pose and Expression-Robust 3D Face Recognition Using Elastic Radial Curves <u>Submitted</u> to **IEEE Transactions on Pattern Analysis and Machine Intelligence.**