

Title: Shape analysis of local facial patches for 3D facial expression recognition

JavaScript is disabled on your browser. Please enable JavaScript to use all the features on this page.

[Skip to main content](#)[Skip to article](#)

ScienceDirect

* Journals & Books

* Help

* Search

My account

Sign in

* Access through ****your organization****

* Purchase PDF

Search ScienceDirect

Article preview

* Abstract

* Introduction

* Section snippets

* References (20)

* Cited by (74)

Pattern Recognition

Volume 44, Issue 8, August 2011, Pages 1581-1589

Shape analysis of local facial patches for 3D facial expression recognition

Author links open overlay panelAhmed Maalej a b, Boulbaba Ben Amor a b, Mohamed Daoudi a b, Anuj Srivastava c, Stefano Berretti d

Show more

Add to Mendeley

Share

Cite

<https://doi.org/10.1016/j.patcog.2011.02.012>Get rights and content

Abstract

In this paper we address the problem of 3D facial expression recognition. We propose a local geometric shape analysis of facial surfaces coupled with machine learning techniques for expression classification. A computation of the length of the geodesic path between corresponding patches, using a Riemannian framework, in a shape space provides a quantitative information about their similarities. These measures are then used as inputs to several classification methods. The experimental results demonstrate the effectiveness of the proposed approach. Using multiboosting and support vector machines (SVM) classifiers, we achieved 98.81% and 97.75% recognition average rates, respectively, for recognition of the six prototypical facial expressions on BU-3DFE database. A comparative study using the same experimental setting shows that the suggested approach outperforms previous work.

Highlights

? We address the 3D facial expression recognition problem using Riemannian geometry. ? We propose local shape analysis of faces coupled with machine-learning techniques. ? A comparative study shows that the suggested approach outperforms previous work.

Introduction

In recent years, 3D facial expression recognition has received growing attention. It has become an active research topic in computer vision and

pattern recognition community, impacting important applications in fields related to human-machine interaction (e.g., interactive computer games) and psychological research. Increasing attention has been given to 3D acquisition systems due to the natural fascination induced by 3D objects visualization and rendering. In addition 3D data have advantages over the 2D data, in that 3D facial data have high resolution and convey valuable information that overcomes the problem of pose/lighting variations and the detail concealment of low resolution acquisition.

In this paper we present a novel approach for 3D identity-independent facial expression recognition based on a local shape analysis. Unlike the identity recognition task that has been the subject of many papers, only few works have addressed 3D facial expression recognition. This could be explained through the challenge imposed by the demanding security and surveillance requirements. Besides, there has long been a shortage of publicly available 3D facial expression databases that serve the researchers exploring 3D information to understand human behaviors and emotions. The main task is to classify the facial expression of a given 3D model, into one of the six prototypical expressions, namely `_Happiness_`, `_Anger_`, `_Fear_`, `_Disgust_`, `_Sadness_` and `_Surprise_`. It is stated that these expressions are universal among human ethnicity as described in [1], [2].

The remainder of this paper is organized as follows. First, a brief overview of related work is presented in Section 2. In Section 3 we describe the BU-3DFE database designed to explore 3D information and improve facial expression recognition. In Section 4, we summarize the shape analysis framework applied earlier for 3D curves matching by Joshi et al. [3], and discuss its use to perform 3D patches analysis. This framework is further expounded in Section 5, so as to define methods for shapes analysis and matching. In Section 6 a description of the feature vector and used classifiers is given. In Section 7, experiments and results of our approach are reported, and the average recognition rate over 97% is achieved using machine-learning algorithms for the recognition of facial expressions such as multiboosting and SVM. Finally, discussion and conclusion are given in Section 8.

Section snippets

Related work

Facial expression recognition has been extensively studied over the past decades especially in 2D domain (e.g., images and videos) resulting in a valuable enhancement. Existing approaches that address facial expression recognition can be divided into three categories: (1) `_static_` versus `_dynamic_`; (2) `_global_` versus `_local_`; (3) `_2D_` versus `_3D_`. Most of the approaches are based on feature extraction/detection as a mean to represent and understand facial expressions. Pantic and Rothkrantz [4] and Samal

Database description

BU-3DFE is one of the very few publicly available databases of annotated 3D facial expressions, collected by Yin et al. [12] at Binghamton University. It was designed for research on 3D human face and facial expression and to develop a general understanding of the human behavior. Thus the BU-3DFE database is beneficial for several fields and applications dealing with human computer interaction, security, communication, psychology, etc. There are a total of 100 subjects in the database, 56

3D facial patches-based representation

Most of the earlier work in 3D shape analysis use shape descriptors such as curvature, crest lines, shape index (e.g., ridge, saddle, rut, dome, etc.).

These descriptors are defined based on the geometric and topological

properties of the 3D object, and are used as features to simplify the representation and thus the comparison for 3D shape matching and recognition tasks. Despite their rigorous definition, such features are computed based on numerical approximation that involves second

Framework for 3D shape analysis

Once the patches are extracted, we aim at studying their shape and design a similarity measure between corresponding ones on different scans under different expressions. This is motivated by the common belief that people smile, or convey any other expression, the same way, or more appropriately certain regions taking part in a specific expression undergo practically the same dynamical deformation process. We expect that certain corresponding patches associated with the same given expression

Feature vector generation for classification

In order to classify expressions, we build a feature vector for each facial scan. Given a candidate facial scan of a person j , facial patches are extracted around facial landmarks. For a facial patch P_{j_i} , a set of level curves $\{c_j^i\}$ are extracted centered on the i -th landmark. Similarly, a patch P_{ref_i} is extracted in correspondence to landmarks of a reference scans ref . The length of the geodesic path between each level curve and its corresponding curve on the reference scan is computed using a

Recognition experiments

To investigate facial expression recognition, we have applied our proposed approach on a dataset that is appropriate for this task. In this section, we describe the experiments, obtained results and comparisons with related work.

Conclusions

In this paper we presented a novel approach for identity-independent facial expression recognition from 3D facial shapes. Our idea was to describe the change in facial expression as a deformation in the vicinity of facial patches in 3D shape scan. An automatic extraction of local curve-based patches within the 3D facial surfaces was proposed. These patches were used as local shape descriptors for facial expression representation. A Riemannian framework was applied to compute the geodesic path

****Ahmed Maalej**** is currently a Ph.D. candidate within the Fundamental Computer Science laboratory of Lille (LIFL UMR 8022), France. He obtained the M.S. degree in Telecommunications from the Higher School of Communications of Tunis (SUPCOM), Tunisia, in 2008, and the electrical engineering degree from the National Engineering School of Monastir (ENIM), Tunisia, in 2005. His main research interests focus on 3D facial expression recognition.

Recommended articles

References (20)

* A. Samal _et al._

Automatic recognition and analysis of human faces and facial expressions: a survey

Pattern Recognition

(1992)

* P. Ekman, T.S. Huang, T.J. Sejnowski, J.C. Hager, Final report to nsf of the planning workshop on facial expression...

* P. Ekman, W.V. Friesen, Constants Across Cultures in the Face and Emotion,...

* S. Joshi, E. Klassen, A. Srivastava, I.H. Jermyn, A novel representation for Riemannian analysis of elastic curves in...

* M. Pantic _et al._

Automatic analysis of facial expressions: the state of the art

IEEE Transactions on Pattern Analysis and Machine Intelligence

(2000)

* J. Whitehill _et al._

Local versus global segmentation for facial expression recognition

* P. Ekman, W. Friesen, Facial Action Coding System: A Technique for the Measurement of Facial Movement,...

* J. Wang _et al._

3D facial expression recognition based on primitive surface feature distribution

* H. Soyel _et al._

Facial expression recognition using 3d facial feature distances

* I. Mpiperis _et al._

Bilinear models for 3d face and facial expression recognition

IEEE Transactions on Information Forensics and Security
(2008)

There are more references available in the full text version of this article.

Cited by (74)

* ### Facial expression recognition with Convolutional Neural Networks: Coping with few data and the training sample order

2017, Pattern Recognition

Citation Excerpt :

However, in this work, just the 2D image of the subjects are used to recognize the expressions. There are other works in the literature that presents higher facial expression recognition accuracies, but using 3D information to infer the expressions [72?74]. A better evaluation of the proposed method in real environments is the cross-database experiments, i.e. train the method with one database and test with another (in this case the BU-3DFE).

Show abstract

Facial expression recognition has been an active research area in the past 10 years, with growing application areas including avatar animation, neuromarketing and sociable robots. The recognition of facial expressions is not an easy problem for machine learning methods, since people can vary significantly in the way they show their expressions. Even images of the same person in the same facial expression can vary in brightness, background and pose, and these variations are emphasized if considering different subjects (because of variations in shape, ethnicity among others). Although facial expression recognition is very studied in the literature, few works perform fair evaluation avoiding mixing subjects while training and testing the proposed algorithms. Hence, facial expression recognition is still a challenging problem in computer vision. In this work, we propose a simple solution for facial expression recognition that uses a combination of Convolutional Neural Network and specific image pre-processing steps. Convolutional Neural Networks achieve better accuracy with big data. However, there are no publicly available datasets with sufficient data for facial expression recognition with deep architectures. Therefore, to tackle the problem, we apply some pre-processing techniques to extract only expression specific features from a face image and explore the presentation order of the samples during training. The experiments employed to evaluate our technique were carried out using three largely used public databases (CK+, JAFFE and BU-3DFE). A study of the impact of each image pre-processing operation in the accuracy rate is presented. The proposed method: achieves competitive results when compared with other facial expression recognition methods ? 96.76% of accuracy in the CK+ database ? it is fast to train, and it allows for real time facial expression recognition with standard computers.

* ### Static and dynamic 3D facial expression recognition: A comprehensive survey
2012, Image and Vision Computing

Citation Excerpt :

One of the main methods of classification that have been employed is Support Vector Machines (SVMs) [110,99,96,68,101,116,109,95,111], including multi-class SVMs [92,117]. Another technique that has been widely used is AdaBoost classification [106,99,101,109] with a selection of different weak classifiers such as linear regressors and LDA. A variation of this is GentleBoost classification, which was employed in [56,57].

Show abstract

Automatic facial expression recognition constitutes an active research field due to the latest advances in computing technology that make the user's experience a clear priority. The majority of work conducted in this area involves 2D imagery, despite the problems this presents due to inherent pose and illumination variations. In order to deal with these problems, 3D and 4D (dynamic 3D) recordings are increasingly used in expression analysis research. In this paper we survey the recent advances in 3D and 4D facial expression recognition. We discuss developments in 3D facial data acquisition and tracking, and present currently available 3D/4D face databases suitable for 3D/4D facial expressions analysis as well as the existing facial expression recognition systems that exploit either 3D or 4D data in detail. Finally, challenges that have to be addressed if 3D facial expression recognition systems are to become a part of future applications are extensively discussed.

* ### A brief review of facial emotion recognition based on visual information
2018, Sensors (Switzerland)

* ### Multimodal 2D+3D Facial Expression Recognition with Deep Fusion Convolutional Neural Network
2017, IEEE Transactions on Multimedia

* ### Survey on RGB, 3D, Thermal, and Multimodal Approaches for Facial Expression Recognition: History, Trends, and Affect-Related Applications
2016, IEEE Transactions on Pattern Analysis and Machine Intelligence

* ### 3D facial expression recognition using SIFT descriptors of automatically detected keypoints
2011, Visual Computer

View all citing articles on Scopus

****Ahmed Maalej**** is currently a Ph.D. candidate within the Fundamental Computer Science laboratory of Lille (LIFL UMR 8022), France. He obtained the M.S. degree in Telecommunications from the Higher School of Communications of Tunis (SUP?COM), Tunisia, in 2008, and the electrical engineering degree from the National Engineering School of Monastir (ENIM), Tunisia, in 2005. His main research interests focus on 3D facial expression recognition.

****Boulbaba Ben Amor**** received the M.Sc. degree in 2003 and the Ph.D. degree in Computer Science in 2006, both from Ecole Centrale de Lyon, France. Currently, he is an associate-professor in Institut Telecom/Telecom Lille 1. He is also a member of the Computer Science Laboratory in University Lille 1 (LIFL UMR CNRS 8022). His research interests are mainly focused on statistical three-dimensional face analysis and recognition and facial expression recognition using 3D. He is co-author of several papers in refereed journals and proceedings of international conferences. He has been involved in French and International projects and has served as program committee member and reviewer for international journals and conferences.

****Mohamed Daoudi**** is a Full Professor of Computer Science in the Institut TELECOM ; TELECOM Lille 1. He received the Ph.D. degree in Computer Engineering from the University Lille 1, France, in 1993 and Habilitation à Diriger des Recherches (HDR) from the University of Littoral, France, in 2000. He is the founder and the head of the MIIRE research group of LIFL (UMR CNRS 8022). His research interests include pattern recognition, image processing, invariant representation of images and shapes, three-dimensional analysis and

retrieval and more recently 3D face recognition. He has published more than 80 papers in refereed journals and proceedings of international conferences. He is the author the book 3D Processing: Compression, Indexing and Watermarking (Wiley, 2008). He has served as a Program Committee member for the International Conference on Pattern Recognition (ICPR) in 2004 and the International Conference on Multimedia and Expo (ICME) in 2004 and 2005. He is a frequent reviewer for IEEE Transactions on Pattern Analysis and Machine Intelligence and for Pattern Recognition Letters. His research has been funded by ANR, RNRT and European Commission grants. He is Senior Member of IEEE.

****Anuj Srivastava**** is a Professor of Statistics at Florida State University in Tallahassee, FL. He obtained his MS and Ph.D. degrees in Electrical Engineering from Washington University in St. Louis in 1993 and 1996, respectively. After spending the year 1996-1997 at Brown University as a visiting researcher, he joined FSU as an Assistant Professor in 1997. He has received the Developing Scholar and the Graduate Faculty Mentor Awards at FSU. His research is focused on pattern theoretic approaches to problems in image analysis, computer vision, and signal processing. He has developed computational tools for performing statistical inferences on certain non-linear manifolds, in particular the shape spaces of curves and surfaces. He has published over 120 journal and conference articles in these areas.

****Stefano Berretti**** received the Laurea Degree in Electronics Engineering, the Postlaurea Degree in Multimedia Content Design, and the Ph.D. degree in Information and Telecommunications engineering in 1997, 2000, and 2001, respectively, from the University of Florence, Italy, where, since 2002, he has been an assistant professor and teaches "Operating Systems" and "Fundamentals of Computer Programming" in the School of Computer Engineering, and since 2001, he has also been teaching "Database Systems" in the postdoctoral school in "Multimedia Content Design." His scientific interests are pattern recognition, content-based image retrieval, 3D object partitioning and retrieval, and 3D face recognition.

[View full text](#)

Copyright © 2011 Elsevier Ltd. Published by Elsevier B.V. All rights reserved.

Recommended articles

* ### Design of highly active and stable (K_{0.48}Na_{0.48}Li_{0.04})(Nb_{0.975}Sb_{0.025})O₃?(Bi_{0.5}Na_{0.5})(Zr_{0.8}Ti_{0.2})O₃ lead-free piezoelectric ceramics by constructing rhombohedral-tetragonal phase boundaries at room temperature
Ceramics International, Volume 50, Issue 17, Part A, 2024, pp. 29997-30007

Tianzhuo Hu, ?, Mei Zhang

* ### Enhanced electrocaloric response and energy storage in [(Bi_{0.5}Na_{0.5})_{0.94}Ba_{0.06}]_{0.975}Sr_{0.025}TiO₃ ceramic close to room temperature

Radiation Physics and Chemistry, Volume 221, 2024, Article 111765

Issa Kriaa, ?, Ahmed Maalej

* ### Psychological stress of university students in the hardest-hit areas at different stages of the COVID-19 epidemic

Children and Youth Services Review, Volume 125, 2021, Article 105980

Yan Zhang, ?, Fei Huang

* ### Adaptive setpoint temperatures to reduce the risk of energy poverty? A local case study in Seville
Energy and Buildings, Volume 231, 2021, Article 110571

David Bienvenido-Huertas, ?, Carlos Rubio-Bellido

* ### Contribution of (Bi_{0.45}Y_{0.05})Na_{0.5}ZrO₃ to induced multiphase coexistence and enhanced piezoelectric properties of K_{0.5}Na_{0.5}NbO₃ lead-free ceramics

Materials Science and Engineering: B, Volume 305, 2024, Article 117416

Xianzhao Zhang, ?, Zhen Zhang

* ### A facial expression recognition system using robust face features from depth videos and deep learning

Show 3 more articles

Article Metrics

Citations

* Citation Indexes74

Captures

* Readers65

View details

* About ScienceDirect

* Remote access

* Shopping cart

* Advertise

* Contact and support

* Terms and conditions

* Privacy policy

Cookies are used by this site. [Cookie Settings](#)

All content on this site: Copyright © 2024 Elsevier B.V., its licensors, and contributors. All rights are reserved, including those for text and data mining, AI training, and similar technologies. For all open access content, the Creative Commons licensing terms apply.

We use cookies that are necessary to make our site work. We may also use additional cookies to analyze, improve, and personalize our content and your digital experience. For more information, see our [Cookie Policy](#)

[Cookie Settings](#) [Accept all cookies](#)

[Cookie Preference Center](#)

We use cookies which are necessary to make our site work. We may also use additional cookies to analyse, improve and personalise our content and your digital experience. For more information, see our [Cookie Policy](#) and the list of [Google Ad-Tech Vendors](#).

You may choose not to allow some types of cookies. However, blocking some types may impact your experience of our site and the services we are able to offer. See the different category headings below to find out more or change your settings.

[Allow all](#)

[Manage Consent Preferences](#)

[Strictly Necessary Cookies](#)

Always active

These cookies are necessary for the website to function and cannot be switched off in our systems. They are usually only set in response to actions made by you which amount to a request for services, such as setting your privacy preferences, logging in or filling in forms. You can set your browser to block or alert you about these cookies, but some parts of the site will not then work. These cookies do not store any personally identifiable information.

[Cookie Details List?](#)

[Functional Cookies](#)

Functional Cookies

These cookies enable the website to provide enhanced functionality and personalisation. They may be set by us or by third party providers whose services we have added to our pages. If you do not allow these cookies then some or all of these services may not function properly.

[Cookie Details List?](#)

[Performance Cookies](#)

Performance Cookies

These cookies allow us to count visits and traffic sources so we can measure and improve the performance of our site. They help us to know which pages are the most and least popular and see how visitors move around the site.

Cookie Details List?

Targeting Cookies

Targeting Cookies

These cookies may be set through our site by our advertising partners. They may be used by those companies to build a profile of your interests and show you relevant adverts on other sites. If you do not allow these cookies, you will experience less targeted advertising.

Cookie Details List?

Back Button

Cookie List

Search Icon

Filter Icon

Clear

checkbox label label

Apply Cancel

Consent Leg.Interest

checkbox label label

checkbox label label

checkbox label label

Confirm my choices