

Guest Editorial

Introduction to the Special Issue on Recent Advances in Biometric Systems

WE ARE pleased to present 14 papers in this special issue devoted to recent advances in biometric systems. A total of 78 papers were submitted for consideration for the special issue. Those that appear in this special issue result from a careful review process and consideration of timing for the special issue. Other papers, which were originally submitted for consideration for the special issue, may be undergoing major revisions and resubmission and appear at a later time in a regular issue of this journal or possibly in some other journal. In particular, several submissions in the area of iris biometrics could not be considered for this special issue due to their experimental results being based primarily on the CASIA 1 iris image dataset [1].

Papers on a broad variety of topics were submitted to the special issue. The large active areas of biometrics such as face, fingerprint, voice, signature, and iris were naturally well represented in the submissions. Newer and smaller areas such as gait and ear biometrics were also represented. Even more unusual areas such as brain signal recordings and infrared imaging of hand vein patterns were also represented. The diversity of topics in the submitted papers is reflected to some degree in the accepted papers and is an indication of the broad and vibrant current nature of the field.

Security and privacy issues in large biometrics systems have received relatively less attention in the past. We are indeed fortunate to have two excellent papers in this area, dealing with what are called “revocable” or “cancelable” biometrics. The first paper works in the context of face recognition and the second paper models forgery for behavioral biometrics.

The paper “Cancelable Biometrics Realization with Multi-space Random Projections” by Teoh and Yuang addresses both revocability and privacy of biometrics templates using a two-factor cancelable formulation. In the first step, the biometric data are distorted by transforming the raw biometric data into a fixed-length feature vector in a nonreversible but revocable manner. In the second step, the feature vector is rejected onto a sequence of random subspaces derived from user-specific pseudorandom numbers (PRNs). This process is invertible, thus making the replacement of biometrics possible by replacement of the PRNs. The proposed method has been verified using the FERET face database [10].

Ballard *et al.* present a stimulating paper on evaluation methodologies for behavioral biometrics that take into account threat models which have been, thus far, largely ignored. They argue that trained and target-selected forgers (in the framework of a generative attack model) must be considered to accurately assess the true security afforded by a biometric system. While basing the experiments on handwriting modality, they provide a blueprint for carrying out threat assessment of other behavioral biometrics as well.

Often, multibiometrics is viewed as improving security and performance of biometrics systems. We have three interesting papers covering novel research in the area of biometrics fusion. Gait recognition is a novel biometric that received increased visibility in the research community through the “Human ID at a Distance” program [4]. The paper “Integrating Face and Gait for Human Recognition at a Distance in Video” by Zhou and Bhanu represents the latest trend related to this area, which is the multibiometric combination of face and gait. Previous work on this topic has assumed the ideal view for each modality, a side view for gait, and a frontal view for face. Zhou and Bhanu tackle the more practical but also more challenging problem of using the information for both modalities that can be extracted from the same view. They extract both face and gait information from a side view, using an enhanced side face image and a gait energy image, respectively. They report results of experiments involving 100 video sequences from 45 people and compare the performance of the individual biometrics and different fusion methods. This paper should be of interest to all those working on either face recognition or recognition by gait.

Three-dimensional face recognition is an active area of research in recent years [8]. It is touted by many in the biometrics community today as the way to overcome the complaints that 2-D face recognition cannot adequately deal with changes in pose and illumination, and is also vulnerable to spoofing. Lin *et al.* from the University of Wisconsin describe a 3-D face recognition method that considers features from multiple facial regions, in contrast to previous single-region approaches. They use an LDA-based approach to assign weights and perform fusion of features from the different regions. The paper reports significant improvement on the face recognition grand challenge (FRGC) dataset and robustness of the method even in the presence of facial expressions.

The paper “Fusing Face Recognition Algorithms and Humans” by O’Toole *et al.* is another paper that should be of interest to everyone working in the field of face recognition. Comparison of the face recognition abilities of humans and algorithms is a topic of broad interest and importance, one

92 touched on by these same authors in another recent paper [5]
93 and by Adler and Schuckers in this special issue in the paper
94 mentioned below. However, this paper goes beyond comparing
95 the abilities of humans and algorithms to the combination of
96 the abilities of humans and algorithms. This is potentially a
97 very important and useful topic in any system in which there
98 will be a person monitoring or interpreting the results of a
99 biometric algorithm. This paper first looks at fusing the results
100 of algorithms in experiments using data from the FRGC [6] and,
101 then, considers the problem of fusing the results from human
102 and algorithm recognition, with the goal of maximizing face
103 recognition performance through hybrid systems consisting of
104 multiple algorithms and humans.

105 The paper “Individual Kernel Tensor-Subspaces for Robust
106 Face Recognition” by Park and Savvides describes a face recog-
107 nition method that uses tensors (high-order matrices) to extract
108 more information from a single face image than other linear
109 models (such as PCA) by categorizing face images according
110 to each factor, such as people, pose, and illumination, and
111 analyzing the bases of the factor. It proposes an efficient method
112 that does not require tensor factorization for classifying test
113 images. Experimental results are reported on the CMU PIE
114 database.

115 Everyone with an interest in iris biometrics will want to
116 read the paper “New Methods in Iris Recognition” by John
117 Daugman. The development of the iris biometrics field has
118 been heavily influenced by Daugman’s work [2], and this paper
119 presents the latest results in his line of work. The state of the
120 art in iris biometrics algorithms has substantially changed since
121 the beginning of this relatively young field [9]. Whereas circular
122 outlines are assumed to be adequate models of the iris boundary
123 in nearly all of the existing iris biometrics literature, this latest
124 work shows that an improved performance can be gained by
125 going to active contours that allow noncircular boundaries. It
126 also shows how eyelash occlusion of the iris region can be
127 detected using statistical inference, attacks the difficult problem
128 of dealing with off-axis gaze, and discusses score normalization
129 and large-scale databases. Results are presented for the ICE
130 2005 dataset [3] and the UAE dataset.

131 The paper “On Techniques for Angle Compensation in Non-
132 Ideal Iris Recognition” by Schuckers *et al.* attacks a prob-
133 lem in making iris biometrics work in a more flexible user
134 interface. Current commercial iris biometric systems require
135 substantial user cooperation in positioning the eye for image
136 acquisition, with the goal of obtaining a good quality image
137 from an approximately frontal view. This paper focuses on
138 techniques for dealing with a particular type of nonideal image,
139 one that is acquired from an off-angle, rather than a frontal view.
140 This is an important topic that will undoubtedly see more
141 activity in the near future.

142 Despite decades of research in fingerprint recognition, many
143 challenges still exist. The paper “Fingerprint Image Mosaick-
144 ing by Recursive Ridge Mapping” by Choi *et al.* deals with
145 the issue of obtaining a larger fingerprint image by stitching
146 together smaller images. Their approach matches ridges itera-
147 tively in order to overcome the problem of correspondences and
148 compensates for the amount of plastic distortion between two
149 partial images by using a thin plate spline model. By using a

three-step process of feature extraction, transform estimation, 150
and mosaicking, the proposed algorithm starts with a trans- 151
form, which is initially estimated with matched minutiae and 152
the attached ridges. Unpaired ridges in the overlapping area 153
between two images are matched iteratively by minimizing 154
the registration cost, which consists of ridge matching error 155
and inverse consistency error. During the estimation, erroneous 156
correspondences are eliminated by considering the geometric 157
relationship between the correspondences and by minimizing 158
the registration cost. The proposed algorithm has been tested 159
on FVC 2002 database [7], and results are compared with three 160
existing approaches to show the usefulness of the proposed 161
approach. 162

Another fingerprint analysis paper “Modeling and Analysis 163
of Local Comprehensive Minutiae Relation for Fingerprint 164
Matching” by He *et al.* describes a graph-based method for 165
fingerprint matching. With the comprehensive minutiae points 166
acting as the vertex set and the local binary minutiae relations 167
providing the edge set, a graph representation of the finger- 168
print is constructed. From the binary relations represented by 169
the edge set, both transformation-invariant and transformation- 170
variant features are extracted. The transformation-invariant fea- 171
tures are used in estimating the local matching probability, 172
while the transform-variant features are used in modeling the 173
fingerprint rotation transformation. The final stage of matching 174
is conducted with a variable bounded-box method and iterative 175
strategy. Experiments that are based on FVC 2002 [7] show 176
that the proposed scheme is effective and robust in terms of 177
fingerprint alignment and matching. 178

Many approaches have been proposed to improve face recog- 179
nition performance that can tolerate pose variations. The paper 180
“A Mosaicing Scheme for Pose-Invariant Face Recognition” by 181
Singh *et al.* proposes a scheme to generate a composite face 182
image during enrollment based on the evidence provided by 183
frontal and semiprofile face images of an individual, obviating 184
the need to store multiple face templates representing multiple 185
poses. A composite face image is computed using multiresolu- 186
tion splining to blend the side profiles with the frontal image. 187
Experiments conducted on three different databases using a 188
texture-based face recognition engine (a modified version of the 189
C2 algorithm) indicate significant benefits of the proposed face 190
mosaicing scheme in improving recognition performance in 191
the midst of pose variations. 192

Machine learning researchers will find the face recognition 193
paper by Xu *et al.* extremely interesting. It deals with repre- 194
sentation of high-dimensional face data as tensors to reduce 195
the parameters that must be learned. Given the perpetual 196
problem of insufficient training data, dimensionality reduction 197
by tensor representation has recently gained popularity. The 198
authors show that the supervised subspace learning algorithm, 199
rank-one projections and adaptive margins, or RPAM, offers 200
many advantages over other dimensionality reduction methods 201
and reports promising numbers on the CMU PIE dataset. 202

Signature verification advances are described by Van *et al.* 203
in a comprehensive experimental evaluation on the SVC2004, 204
BIOMET, PHILIPS, and MYCT datasets. They introduce the 205
notion of a “segmentation information” score that is derived by 206
analyzing the Viterbi path, which is then fused with the hidden 207

208 Markov model (HMM) likelihood score. This is an interesting
209 and novel approach as both scores are generated by the HMM
210 for each writer. The paper also describes a sophisticated per-
211 sonal normalization scheme that is reported to hold up well
212 across the datasets.

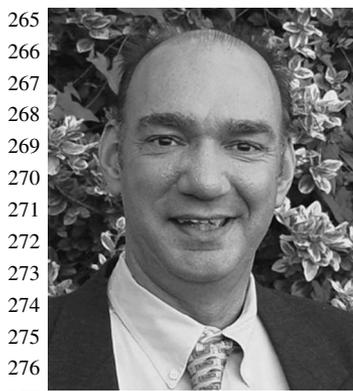
213 The paper “Comparing Human and Automatic Face Rec-
214 ognition Performance” by Adler and Schuckers contains several
215 elements that will be of broad interest to the face recog-
216 nition community. Table I of their paper tracks a compar-
217 ison of human and automatic face recognition performance
218 from 1999 to 2006. It shows a pattern where human face
219 recognition started out performing much better than automatic
220 face recognition, but automatic recognition improved over time
221 to the point where it now outperforms human face recognition.
222 Those who find this result interesting and/or controversial
223 will want to examine, in more detail, the methodology un-
224 derlying the result. They also present a new methodology to
225 calculate an average detection error tradeoff (DET) curve. The
226 DET curve is related to the receiver operating characteristic
227 curve.

228 We want to thank the authors, the reviewers, and the Trans-
229 actions staff for all of the effort that has gone into producing
230 this special issue. We feel confident that the reader will see the
231 fruits of this effort in the many interesting, challenging, and
232 surprising results presented in the papers in this Special Issue.

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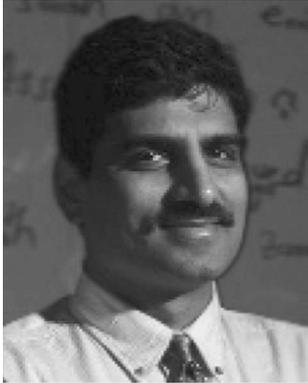


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He is currently the Schubmehl-Prein Professor and Chair of the Department of Computer Science and Engineering, University of Notre Dame, Notre Dame, IN. He was a North American Editor of the *Image and Vision Computing Journal*. He also serves or has served on the editorial boards of *Computer Vision and Image Understanding*, *Machine Vision and Applications*, *International Journal of Pattern Recognition and Artificial Intelligence*, *Pattern Recognition*, *Electronic Letters in Computer Vision and Image Analysis*, and *Journal of Privacy Technology*. His current research interests are data mining and biometrics. The data mining research, which is aimed at ensemble methods for “extreme” problems, has been supported by Sandia National Laboratories, Albuquerque, NM. The biometrics research has been supported by a number of agencies, and the Notre Dame Research Group has been active in support of the government’s Face Recognition Grand Challenge Program and the Iris Challenge Evaluation Program.

277 Prof. Bowyer is a Golden Core Member of the IEEE Computer Society. He received the Award
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282 Theory, Applications and Systems (BTAS 2007).

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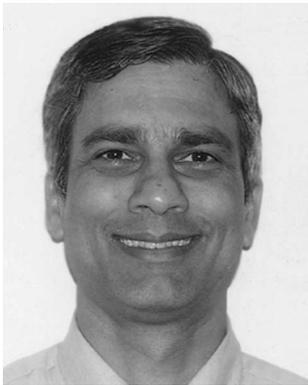


Venu Govindaraju.

283 AQ2

He founded the Center for Unified Biometrics and Sensors, Amherst, NY in 2003, which 284 has received over \$5 million in research funding covering over a dozen projects from both the 285 government and the industry. He has supervised the dissertation of ten doctoral students and 286 the theses of over 20 Master's students. He is currently a Professor of computer science and 287 engineering with the Department of Computer Science and Engineering, University at Buffalo, 288 The State University of New York, Amherst. In a research career spanning over 20 years, he has 289 made significant contributions to pattern recognition such as document analysis and biometrics. 290 His seminal work in handwriting recognition was at the core of the first handwritten address 291 interpretation system used by the U.S. Postal Service. 292

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296 AQ3

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