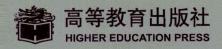
Dengpan Mou

# Machine-based Intelligent Face Recognition

基于机器的智能人脸识别



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With 58 figures



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#### **Preface**

We can't solve the problems by using the same kind of thinking we used when we created them.

Albert Einstein (1879-1955)

State-of-the-art machine-based face recognition technology, although booming since last decades, is still suffering a lot from critical research challenges, such as the lack of fundamental intelligence, the difficulties of running completely automatically and unsupervisedly without separate training, and the typical failures of dealing with free face pose variations, etc. Those limitations greatly hinder the wide applications it could have had. This book is the first to discuss the general engineering methods of imitating intelligent human brains for video-based face recognition. The advances and evidences from the cognitive science research are introduced in this book, which further strengthen our thoughts and proposals to achieve such a fundamental intelligence in machine vision.

Regarding intelligence, we have defined two directions. The first effort is to simulate the ability of self-learning, self-matching and self-updating. This side of intelligence can be detailed into the following features: the whole recognition procedure is running in an unsupervised, automatic, non-invasive, and self-updated way. It is important to note that, the fully automatic procedure is a generalized face recognition procedure, which includes the task of enrollment (training) and updating as well. However, those steps are typically separate and supervised in machine learning, and therefore missing the essentials of intelligence.

The other main focus of the book is to explore the novel ways on how to implement the high-level analysis in machine-based face recognition, to simu-

#### VI Preface

late the process in human brains. Through high-level analysis, it is possible to combine multiple available methods, which include conventional machine learning algorithms, image processing approaches, predefined rules, video context, temporal and spatial correlations and even logic deduction. The fusion of multiple approaches contributes significantly to the improved face recognition performance.

Experiments are made through long-term (over years) constructed sequences with more than 30 specific subjects and more than 20 faces from TV news channels. The evaluation results demonstrate the robustness of the proposals in unconstraint video scenarios.

The objective of the author is to provide this book for scientists, researchers and students in the areas of machine-based face recognition. The fundamentals and research backgrounds are provided, aiming to help the beginners to quickly step into the field. Introduction and analysis of the state-of-the-art technology can assist experts to easily keep up to date with the world-wide overview.

The author does hope that the proposals to achieve both intelligence and robustness could be somehow helpful for other researchers, to finally popularize the technology, and to pervasively apply it for designing general machine learning algorithms.

Dengpan Mou Villingen-Schwenningen, Germany April 2009

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In writing this book, I have been fortunate to benefit from valuable research discussions with a large number of research groups through my technical reports in many different universities and conferences during the last several years. Although there is no space to mention the hundreds of researchers and scientists for their helpful comments, I would like to take this opportunity to generally thank them all.

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# **Contents**

1	Introdu	ction	1
	1.1 Fac	e Recognition—Machine Versus Human	1
		posed Approach	
	1.3 Pro	spective Applications	6
	1.3.1	Recognition in the Future Intelligent Home	6
	1.3.2	Automotive	
	1.3.3	Mobile Phone for Children	9
	1.4 Ou	tlinetline	9
	Reference	es	10
2	Fundan	nentals and Advances in Biometrics and Face Recognition	n13
	2.1 Ger	neralized Biometric Recognition	13
	2.2 Co	gnitive-based Biometric Recognition	16
	2.2.1	Introduction	16
	2.2.2	History of Cognitive Science	17
	2.2.3	Human Brain Structure	18
	2.2.4	Generic Methods in Cognitive Science	
	2.2.5	Visual Function in Human Brain	22
	2.2.6	General Cognitive-based Object Recognition	23
	2.2.7	Cognitive-based Face Recognition	24
	2.2.8	Inspirations from Cognitive-based Face Recognition	28
	2.3 Ma	chine-based Biometric Recognition	
	2.3.1	Introduction	29
	2.3.2	Biometric Recognition Tasks	29
	2.3.3	Enrollment—a Special Biometric Procedure	30
	2.3.4	Biometric Methods Overview	31
	2.3.5	Fingerprint Recognition	33
	2.4 Ge	neralized Face Recognition Procedure	36
	2.5 Ma	chine-based Face Detection	37
	2.5.1	Face Detection Categories	37
	2.6 Ma	chine-based Face Tracking	39
	2.7 Ma	chine-based Face Recognition	41
	2.7.1	Overview	41

* *	C
X	Contents

	2.7.2	Benchmark Studies of Face Recognition	12
	2.7.2	Some General Terms Used in Face Recognition	
	2.7.4	Recognition Procedures and Methods	
	2.7.5	Video-based Recognition	
	2.7.6	Unsupervised and Fully Automatic Approaches	
		nmary and Discussions	
	Referenc	es	61
3	Combin	ed Face Detection and Tracking Methods	71
J		oduction	
		age-based Face Detection	
		Choice of the Detection Algorithm	
	3.2.1		
	3.2.2	Overview of the Detection Algorithm	
	3.2.3	Face Region Estimation	
	3.2.4	Face Detection Quality	
		nporal-based Face Detection	
	3.3.1	Overview	
	3.3.2		
	3.3.3	Analysis of Temporal Changes	
		nmary	
		ther Discussions	
	Referenc	es	89
4	Automa	tic Face Recognition	91
7		erview	
		ture Extraction and Encoding	
		tching/Classification	
		tening/Classification	
	4.3.1		
	4.3.2		90
	4.3.3	Temporal Filtering	98
		mbined Same Face Decision Algorithms	
	4.5 Sur	nmary	106
	Reference	es	106
5	Unsune	rvised Face Database Construction	107
J	5.1 Inti	roduction	107
	5.2 Bac	ckgrounds for Constructing Face Databases	108
	5.2.1	Supervised Learning	
		Unsupervised Learning	
	5.2.2		
	5.2.3	Clustering Analysis	
		tabase Structure	
	5.3.1	A Fused Clustering Method	
	5.3.2	Parameters in the Proposed Structure	
	5.4 Fea	atures of an Optimum Database	122

Index ......165

# **List of Figures**

Fig. 1.1	Overview of the System Functional Blocks	4
Fig. 1.2	Conceptual Architecture of an Intelligent Home System	6
Fig. 2.1	Topological division of neocortex [8], with changes	20
Fig. 2.2	Market share of biometric techniques [38]	32
Fig. 2.3	Category of fingerprint recognition applications	35
Fig. 2.4	Categories of face detection methods	37
Fig. 2.5	Category of face recognition procedures and methods	46
Fig. 3.1	Function blocks of face detection	73
Fig. 3.2	Proportional illustration of face region depending on the eye	
	distance	75
Fig. 3.3	Example of face extraction	76
Fig. 3.4	Variance of 3D head poses	77
Fig. 3.5	Function blocks of temporal-based face detector	79
Fig. 3.6	Two examples of face region fast movement	80
Fig. 3.7	Proportional illustration of face region and search region	83
Fig. 3.8	Example images to find the range of $T_{\rm M}$	85
Fig. 3.9	Percent of the motion pixels in the search region (face region	
	excluded)	86
Fig. 3.10	Sequences of multiple people	88
Fig. 3.11	1 Proposal for the multiple IFD	88
Fig. 4.1	Feature extraction of a face region	93
Fig. 4.2	Comparison between Bayesian distance and center distance	94
Fig. 4.3	FAR/FRR curves	96
Fig. 4.4	FAR/FRR curves—enrollment from 1 vs. 16	97
Fig. 4.5	Two examples of false positives in face detection	99
Fig. 4.6	An example of face occlusion with slow motion	102
Fig. 5.1	Four different ways of distance measure	110
Fig. 5.2	Sketches of the two clustering methods	112
	<b>,</b>	

Fig.	5.3	Two structures of a face database	114
Fig.	5.4	An example of different people having a big similarity	. 115
Fig.	5.5	Face database structure in two different states	117
Fig.	5.6	Two classes in the testing set S2	120
Fig.	5.7	Examples of the testing set S1	. 121
Fig.	6.1	Definition of all possible states for the face recognition procedure	. 126
Fig.	6.2	Definition with a hierarchical state machine	.128
Fig.	7.1	Hardware configurations of the system	.130
Fig.	7.2	Screenshot of the system running live	.132
Fig.	7.3	System implementation components	.134
Fig.	7.4	Samples of whole frame images and the corresponding face images	.137
Fig.	7.5	Cross-similarity differences by using face images and	
		whole images	.137
Fig.	7.6	Auto-similarity differences by using face images and whole images	. 138
Fig.	8.1	Test of 3D head motions—head roll	.143
Fig.	8.2	Test of 3D head motions—head yaw	. 143
Fig.	8.3	Test of 3D head motions—head pitch	. 144
Fig.	8.4	Sequence of one person with significant 3D head motion	. 145
Fig.	8.5	Performance comparison between an IFD and our combined	
		detector	. 145
Fig.	8.6	Sequence of one person with intentionally facial expression	
		changes	. 146
Fig.	<b>8.7</b>	Sequence of one person with significant scale changes	. 147
Fig.	8.8	Sequence of one person with significant lighting changes	. 148
Fig.	8.9	Sequence of one person with fast motion	. 149
Fig.	8.10	Face recognition test—different face scales	. 150
Fig.	8.11	Face recognition test—different lightings and facial expressions	.151
Fig.	8.12	Face recognition test—Occlusions and with/without glasses	. 152
Fig.	8.13	Face recognition performance test—yaw angle variance	. 153
Fig.	8.14	Examples of artifacts around the eyes which cause recognition	
		errors	. 154
_		Definition of roll angle	. 155
Fig.	8.16	Roll angle influence on the recognition classifier	. 156
Fig.	8.17	Variety and rapidity of database construction	. 157
Fig.	8.18	Updatability and uniqueness of database construction	. 158
Fig.	8.19	Offline performance of the whole system	.160

# **List of Tables**

Table 2.1	Biometric procedures and categories	15
Table 2.2	Comparison between different biometric recognition techniques	32
Table 3.1	Comparison of $M_F$ and actual speed shown in Fig. 3.6	81
Table 4.1	Similarity comparison of two similar persons	. 100
Table 4.2	List of all cases for combined same face decision algorithms	. 104
Table 8.1	Performance comparison between the IFD and our detection	
	method	. 149

#### 1 Introduction

Abstract Face recognition is quite intuitive for most human beings while rather too complex for a machine vision system to be pervasively applied. As the starting point of this book, we are discussing the fundamental pros and cons of state-of-the-art machine vision system versus its human being counterpart. Inspired from the comparison, in the second part of this chapter, a machine-based intelligent face recognition system is briefly introduced. With the high intelligence and high recognition robustness, the system could be widely used for prospective applications such as future smart home, automotive and children mobile phones. In the third part, the discussion of those applications is aiming at broading the visions of the readers. Finally, the outline of the rest part of this book is given.

### 1.1 Face Recognition—Machine Versus Human

Face recognition has been studied for many years and its attraction grows rapidly due to the wide and promising applications, such as daily identification systems (e.g. automatic banking, access control, computer log-in, etc.), in communication systems (e.g. teleconference and video-phone, etc.), in public security systems (e.g. criminal identification, digital driver license, etc.) and in law enforcement systems.

Face recognition seems instinctive for human beings, but it is really a tough and complex task for a machine-based system. Over the decades, scientists in cognitive and neuroscience are always trying to explore how human

#### 2 1 Introduction

beings recognize faces and why we are in general good at recognizing faces. The recent advances of their research are enlightening [1], but their contributions to the mathematical models and engineering solutions for a machine vision system are still far from enough. Therefore, researchers from the computer science are constructing vast numbers of mathematical models and dedicated algorithms, which may cover the field of artificial intelligence, machine learning, image processing and even video signal processing, etc.

Machine vision systems could have several major advantages over humans. It can have a huge storage medium to deal with much larger amount of people. Furthermore, a machine is not easy to be tired and can work 24 hours a day without any problems. It is always expected to replace the human resources to significantly lower the cost. More importantly, the use of machine can keep the privacy. It is always fair and can 100% follow the predefined rules. For example, suppose that a completely automatic system is required to alarm strangers. A machine will not keep the information of known people and therefore is not violating their privacy. But a human supervisor cannot 100% objectively manage it during the monitoring. And in this case, probably all people under supervision would prefer a machine than a human supervisor.

The most recent face recognition tests [1, 2] claim that, in a particular case, when non-familiar faces (frontal) are tested with significant illumination variations, machine-based face recognition algorithms could be superior to human brains. Nevertheless, in terms of the general recognition precision, the most successful state-of-the-art face recognition technology is still not able to compete with the level of human systems. From the engineering point of view, there is too little attention to the research on face recognition by imitating a human being in a fundamental way, which could combine every means for recognition, not necessarily based on pure psychophysical/neurobiological science or pure mathematical models. For example, it is just a piece of cake for a nine-year old child to recognize people when they turn their heads from frontal to profile views in a video sequence, but can lead to failures for most of current face recognition systems. The reason behind it might be that, in this special case, a child applies his/her multiple approaches including "image processing", video context, logic deduction, experiences, etc. to recognize the people, while those machine-based systems are merely using the mathematical-based image processing methods to calculate the correlation.

Another crucial drawback of most current machine recognition tech-

niques is their lack of intelligence. Many systems are not able to memorize the faces by themselves without the help of a human supervisor or the cooperation of the users. Any new user is supposed to follow some instructions to be enrolled into the databases. The instructions are normally from a human supervisor who is also required to select and update the databases. They are hence invasive to the users and are difficult to be applied in many areas like consumer electronics. Quite recently, there are some advances [3, 4, 5] in the research of building an automatic face recognition system. But their assumptions restrict the real applications. The most critical one is that they normally assume only one person existing in a video sequence, which greatly decreases the complexity of the automatic procedure. With the same previous example, the child has no difficulty in self-learning and memorizing unknown faces, identifying known faces even if several people existing with free behaviors. But that would make a typical error for those systems.

Robustness is the most concerned question for the researchers. Recent face recognition surveys [6, 7] reveal that lighting changes, indoor/outdoor changes, pose variations and elapsed time databases are the critical parameters which greatly influence the performance of a face recognition system. But the effect from those parameters is significantly database dependent. If a database has already enrolled different mugshots under various environments and can update with recent views, state-of-the-art face recognition techniques can produce robust enough results.

Inspired from the above, we define an ideal machine-based face recognition system, which is unique due to the following features:

- Self-learning: completely automatic and unsupervised;
- Non-invasive;
- Robust: in unconstrained environment, against pose and lighting changes, occlusion and aging problems.

In this book, we are exploring the ways and algorithms to approach such an ideal system, mainly but not limited to the application of consumer electronics.

#### 1.2 Proposed Approach

The rule-based algorithms for computer vision and pattern recognition used to

#### 4 1 Introduction

be popular more than 15 years ago. For example, to detect faces, people used eyes, noses, mouths and other features to define the corresponding rules. When there are several facial features detected, and the location of the features agrees with some predefined rules, a face is expected to exist. Although intuitive, these rules are not robust enough to deal with partial occlusions, pose and other frequent day-to-day face changes when applied in images. Learn-based methods therefore appeal more attractions. Nearly all the best face detection and recognition algorithms from the state-of-the-art are learn-based. However, the machine learning method generally suffers from requiring huge numbers of examples for training which makes it difficult to build an automatic system without any human supervision or user's cooperation. In this book, we explore the ways to combine some general rules as major methods together with learn-based methods to achieve such a system for face recognition in video.

For the convenience of describing the proposed approaches, we list the overview of the functional blocks of the automatic system [8] in Fig. 1.1.

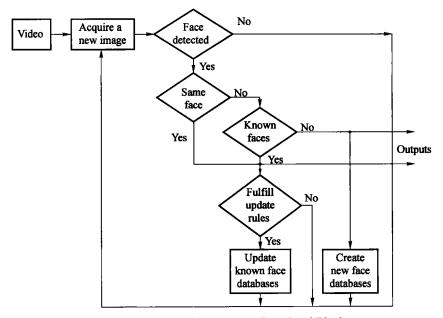


Fig. 1.1 Overview of the System Functional Blocks

Images are continuously acquired from a video source. An image-based face detector and a novel temporal-based face tracker are included to detect whether there are faces in the current image. Any typical image-based face de-