An Overview of Multimodal Biometrics using Meta-Heuristic Optimization Techniques for F2R System

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Abstract—Multimodal biometrics is the combination of more than one unimodal biometrics which involves more accuracy by implementing F2R(Face and Fingerprint Recognition). This survey paper deals with the review of multimodal biometrics using F2R and analyzing various meta heuristic optimization algorithms used at the feature selection level of F2R.

Index Terms — F2R - Face and Fingerprint Recognition, multimodal biometrics, Meta heuristics, Optimization.

I. INTRODUCTION

A biometric system is usually used for recognition based on physiological or behavioral characteristics of an individual. Biometrics system becomes robust method for authentication [1].Biometrics can be classified as Finger print, Face, Knuckle print, palm print, Iris are comes under physiological characteristics whereas Gait, signature and voice are comes under behavioral characteristics. Uniqueness, universality and permanence are the three important aspects in biometrics system.[2].

II. MULTIMODAL BIOMETRICS

Biometrics systems are divided into two categories namely unimodal systems and multimodal systems. Unimodal systems uses only one biometric whereas the multimodal systems uses more than one biometric for identification and verification.[2]. It is commonly used in various engineering applications, border security and immigration purposes and financial applications like ATM withdrawal and fund transfer. The use of multimodal biometrics rather than unimodal biometric will lead to more security and robustness.

A. Multimodal vs UnimodalBiometrics

The advantages of Multimodal biometrics over Unimodal biometric is listed below

- More security
- More Accuracy
- **❖** Anti spoofing
- Universality
- Noisy data resistance
- More Robustness

III. F2R SYSTEM

Most of the successful commercial biometric systems currently rely on either Fingerprint or Facerecognition[3]. So, the Face Recognition and Fingerprint Recognition are taken into account on behalf of multimodal biometric system. The

Revised Version Manuscript Received on October 24, 2015.

Retrieval Number: E2739115515/2015@BEIESP

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F2R system which involves three stages of implementation. They are i)Face Recognition ii) Fingerprint Recognition and iii) Fusion Method.

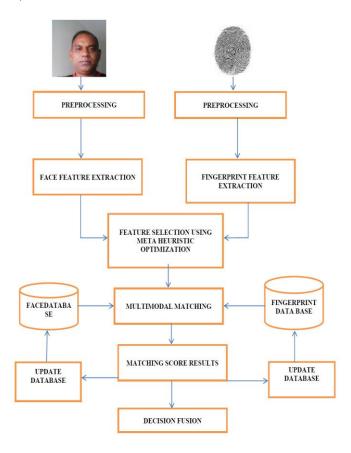


Fig. 1 General framework for F2R system

A. Face Recognition System

Table 1. Summary of Various Face Recognition Techniques

.Methods	Approaches		
	Principal component Analysis[PCA] eigenfaces		
1.Holistic Methods	Probabilistic eigenfaces		
	Fisherfaces/subspace LDA		
	SVM		
	Evolution pursuit		
	Linear Discriminant Analysis(LDA)		
	Probabilistic Decision Based Neural		
	Network(PDBNN)		
	Pure Geometry Methods		
2.Feature Based Methods	Dynamic Link architecture		
	Hidden Markov Model(HMM)		
	Convolution neural network		
	Modular eigenfaces		



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	Hybrid LFA(Local Feature Method)	
Hybrid methods	Shape normalized	
	Component based	

Face recognition comes with three technical approaches, holistic approach, Feature-based approach and hybrid methods[4].

B. Fingerprint Recognition System

Fingerprint recognition comes with three technical approaches, correlation based matching, minutiae matching and non-minutiae feature based matching [5].

Table 2. Summary of Various Fingerprint Recognition Techniques

Methods	Approaches	
	Fourier Transform	
1.Correlation-based	Fourier-Mellin transform	
Matching	Correlation filters	
	Similarity score	
	Point pattern matching	
2.Minutiae based	Simple algebraic geometry	
matching	Hough transform based approach	
	Minutiae matching with	
	pre-alignment	
	Intrinsic co-ordinate system	
	Global and local texture	
	information	
3.Non-minutiae	Geometrical attributes and	
feature based	spatial relationship of the ridge	
matching	lines	
	Level 3 features	

C. Fusion Techniques

Fusion can be accomplished at various levels in a biometric system. Most multi biometric systems fuse information at the match score level or at the decision level.[10] various types of fusion techniques are summarized[5].

Table 3. summary of various fusion techniques

Information level	Fusion Techniques		
Image (or) sensor	Image mosaicking, Image compositing		
Feature	Feature concatenation, template consolidation, template adaptation		
	Sum, mean, median, product, min, max		
Score	Linear discriminant function, logistic regression, neural networks, quadratic classifiers, <i>k</i> -nearest neighbor, decision trees, support vector machines		
	Generalized ensemble, adaptive weighting, stacking, mixture of local experts (MLE), bagging, boosting, random subspace		
Rank	Highest rank, Borda count, weighted Borda count, Dempster-Shafer		
Decision	AND, OR, voting, weighted voting		

IV. META HEURISTIC OPTIMIZATION

Heuristic is a technique that improves the efficiency of the search process. Metaheuristics are widely recognized as efficient approaches for many hard optimization problems especially for Biometric authentication techniques. The taxonomy of various meta heuristic optimization algorithms are listed .[6]

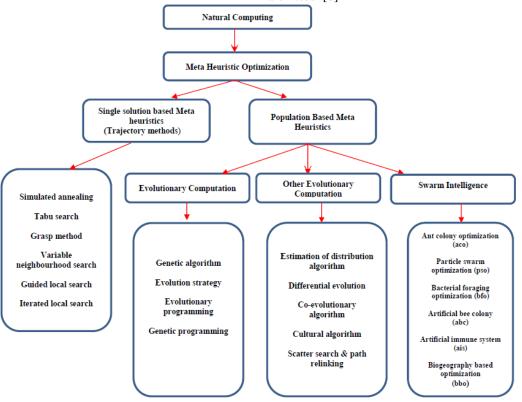


Fig. 2 Taxonomy for Metaheuristic Optimization Techniques



V. COMPARISON OF META HEURISTIC OPTIMIZATION ALGORITHMS

The various meta heuristic optimization algorithms that can be used for Face recognition system and Fingerprint recognition system at the feature selection level, are summarized based on their speed and accuracy levels. [7]

Table 4. Comparison of Accuracy and Speed of Meta Heuristic Algorithms Using the ORL Database

Algorithms	Accuracy	Speed (Sec)	Reference
GA	86.17%	1.334	Li et al. [2010]
GP	63.5%	_	Bozorgtabar et
			al. [2010]
DE	63.12%	6.4	Mallipeddi and
			Lee [2012]
ABC	99%	160	Tran and
			Liatsis [2011b]
PSO	94.5%	160	Cheng et al.
			[2011]
BFO	100%	272.1	Jakhar et al.
			[2011]
ACO	96%	960	Kanan et al.
			[2007]

VI. PERFORMANCE MEASURE

The performance of Face and Fingerprint recognition is measured by two error rates ,False Rejection Rate (FRR) and False Acceptance Rate (FAR).

1) False acceptance rate (FAR), which is defined as the probability of an impostor being accepted as agenuine individual. It is measured as the fraction of impostor score exceeding the predefined threshold. [8,9].

FAR (%) =no. of accepted attempts by imposter X100% total no. of attemptsby impostor

2) False rejection rate (FRR), which is defined as the probability of a genuine individual being rejected as an impostor. It is measured as the fraction of genuinescore below the predefined threshold.[8,9]

FRR (%) =no. of failed attempts by authorized user *X* 100% total no. of attempts by authorized user

VII. CONCLUSION

In this paper, we reviewed the multimodal biometrics system using Face and Fingerprint recognition with various implementation techniques and fusion techniques at various levels. The impact of meta heuristic algorithms takes place at the feature selection level to enhance the performance. Since multimodal biometrics is the emerging technology in the biometric authentication, deploying meta heuristic techniques will effect in high performance and less computational complexity to solve engineering and commercial problems. We reveal that from the survey using the hybrid meta heuristic optimization algorithm will lead to one step ahead to achieve more performance, speed and accuracy than the meta heuristic optimization algorithm.

APPENDIX

F2R- FACE AND FINGERPRINT RECOGNITION

PCA – PRINCIPAL COMPONENT ANALYSIS

LDA – LINEAR DISCRIMINANT ANALYSIS

HMM-HIDDEN MARKOV MODEL

SVM - SUPPORT VECTOR MACHINE

PDBNN- PROBABILISTIC DECISION BASED NEURAL NETWORK

LFA – LOCAL FEATURE METHOD

GA - GENETIC ALGORITHM

DE - DIFFERENTIAL EVOLUTION

GP - GENETIC PROGRAMMING

AIS - ARTIFICIAL IMMUNE SYSTEM

ACO - ANT COLONY OPTIMIZATION

PSO - PARTICLE SWARM OPTIMIZATION

ABC - ARTIFICIAL BEE COLONY

BFO - BACTERIAL FORAGING OPTIMIZATION

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