

FEATURE FUSION USING A MODIFIED GENETIC ALGORITHM FOR FACE
AND SIGNATURE RECOGNITION SYSTEM

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LIST OF SYMBOLS

<i>if-then</i>	-	Rules derived from decision tree
$\{X_1, X_2, X_3, \dots, X_N\}$	-	Training set
N	-	Number of all samples
N_i	-	Number of samples in each class
n -dimensional	-	Dimensional in image space
m -dimensional	-	Dimensional in feature space
$W_{PCA} \in R^{n \times m}$	-	A linear mapping in Eigen space that consist of the highest variance
μ	-	Mean of all samples
μ_i	-	Mean of samples in each class
W_{PCA}	-	Eigen subspace
$W^T S_T W$	-	Scatter matrix of the transformed feature vector
W_{opt}	-	Convex optimization in Linear Discriminant Analysis
S_T	-	The total scatter matrix of samples
$[w_1 \ w_2 \ \dots \ w_g]$	-	The set of n -dimensional eigenvectors of S_T corresponding to the g largest eigenvalues
W_{LDA}	-	Mapping direction matrix in Linear Discriminant Analysis
X'	-	Face protocol derived from a mapping direction matrix in Linear Discriminant Analysis
C	-	Number of classes
S_b	-	Scatter matrix between-class

S_w	- Scatter matrix within-class
Y	- Samples' label
g	- The highest eigenvalues
G	- Pooled within group covariance matrix
x'	- Normalized value
x	- Original value
x_{\min}	- Minimum point of original x point
x_{\max}	- Maximum point of original x point
σ	- Standard deviation of the features
$pbest$	- The best position on i^{th} particle in Particle Swarm Optimization
$gbest$	- The best position of the particles in swarm in Particle Swarm Optimization
X coordinate	- X coordinate of signature writing
Y coordinate	- Y coordinate of signature writing
x_i^o	- Original x point of signature
y_i^o	- Original y point of signature
x_i'	- x point after normalization
y_i'	- y point after normalization
y_{\min}	- Minimum point of original y point
W	- Width of normalized signature data
H	- Height of normalized signature data
Δp	- A fraction of the total arc length of signature data
L	- Arc length of signature data
d	- Distance of point to point in signature data
n	- The number of points in raw signature data
n_1	- Fixed number points in signature data
ΔT	- A fraction of the total time signature
$T(n)$	- Total time of signature
n_2	- Determined fixed number of points after re-sampling process
θ	- Direction of any point in signature

ϕ	- Curvature of any point in signature
$\sin \theta_x(n)$	- Direction of x in local time based signature writing
$\sin \theta_y(n)$	- Direction of y in local time based signature writing
$\cos \theta(n)$	- Direction of x in local strokes based signature writing
$\sin \theta(n)$	- Direction of y in local strokes based signature writing
$\cos \phi(n)$	- Curvature's angle of the strokes between two elementary segment in signature writing
$\sin \phi(n)$	- Curvature's angle of the strokes between two elementary segment in signature writing
ρ_c	- Rate of crossover in GA
ρ_m	- Rate of mutation in GA
ρ_x	- Probability of solution in roulette wheel selection in GA
f_x	- Fitness value for the particular solution in roulette wheel selection in GA
f_{sum}	- Sum of the fitness for all solutions in roulette wheel selection in GA
function fc	- Additional function in GA's fitness function to control the balance of selected features
function fx	- Fitness function which represent the accuracy performance

LIST OF ABBREVIATIONS

GA	-	Genetic Algorithm
LDA	-	Linear Discriminant Analysis
ORL	-	Olivetti Research Laboratory
SUSIG	-	Sabancı University Signature Database
2D	-	Two Dimensional
3D	-	Three Dimensional
PSO	-	Particle Swam Optimization
SVM	-	Support Vector Machine
Min-Max	-	Minimum and Maximum Normalization
Z-score	-	A Statistical Measurement Of A Score's Relationship To The Mean In A Group Of Scores
LR	-	Likelihood Ratio
NN	-	Neural Network
MDR	-	Minimal Distance Rule
FRGC	-	Face Recognition Grand Challenge
MCYT	-	Ministerio de Ciencia y Tecnologia
LCD	-	Liquid-crystal-display
PCA	-	Principle Component Analysis
GAR	-	Genuine Acceptance Rate
FDA	-	Fisher Discriminant Analysis
EER	-	Equal Error Rate
SIFT	-	Scale Invariant Feature Transform Features
ICA	-	Independent Component Analysis
SFFS	-	Sequential Floating Forward Selection Algorithm

GPR	-	Ground Penetrating Radar
FAR	-	False Acceptance Rate
KDDA	-	Kernel Direct Discriminant Analysis
SFS	-	Sequential Forward Selection (SBS)
SBS	-	Sequential Backward Selection
SBFS	-	Sequential Backward Floating Search
k-NN	-	k-Nearest Neighbour
BIOMET	-	Biometrics Database from TELECOM SudParis
SVC2004	-	Signature Verification Competition 2004
YALE	-	Face Database from Yale University
FERET	-	The Facial Recognition Technology Database
UTSB	-	University of Science and Technology Beijing face database

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ABSTRACT

Combination of multi biometrics at feature level fusion is able to give more accurate classification result. This thesis focuses on the development of feature level fusion of bimodal biometrics system for face and dynamic signature recognition system. The modalities of biometric are used due to the ability to avoid spoof attack since it is difficult for impostor to imitate two different characteristics (behaviour and physical) at the same time. Most existing systems are dealing with feature fusion of the same domain such as image based of fingerprint and face. Thus, there is no issue of incompatible features to be fused compared to the proposed development. Balance of the combined features has not been assessed whereas it is essential to ensure one of the biometrics does not dominate accuracy performance. To overcome the issue of incompatible features to be combined, Wrapper Genetic Algorithm (GA) was implemented as the feature selection algorithm due to its ability to evaluate the features irrespective of which domain by masking the features with bit number. A modified fitness function in Wrapper GA was introduced by adding a function to maintain the balanced of the selected features. Penalty's value was imposed to the function when there is imbalance occurs in the selected features. Therefore, the accuracy performance of this system based on the fitness function that will rely on the percentage of correctly recognized samples and the balanced of selected features. Several approaches and benchmark data were used to validate the effectiveness of the proposed method compared to the unimodal system and normal feature selection method. Results show that the proposed method yield optimal recognition with the highest accuracy of 97.50%. In addition, the importance of both biometrics remains, while maintaining the balance of the selected features.

ABSTRAK

Penggabungan pelbagai biometrik di peringkat gabungan ciri-ciri mampu memberi hasil pengkelasan yang lebih tepat. Tesis ini memberi tumpuan kepada pembangunan sistem pengecaman muka dan tandatangan di peringkat tersebut. Jenis biometrik tersebut digunakan kerana ia mampu mengatasi isu penyamaran iaitu sukar untuk penyamar meniru dua ciri yang berbeza (tingkah laku dan fizikal) pada masa yang sama. Kebanyakan sistem sedia ada menggabungkan ciri-ciri daripada domain yang sama seperti cap jari dan muka yang berasaskan imej. Oleh itu, tidak ada isu ketidakserasian untuk digabungkan berbanding dengan pembangunan yang dicadangkan. Keseimbangan ciri-ciri yang digabung juga tidak dinilai sedangkan ia penting untuk memastikan salah satu proses pengkelasan tidak dimonopoli oleh hanya satu biometrik. Untuk mengatasi isu ketidakserasian, Algoritma Genetik *Wrapper* (GA) digunakan untuk memilih dan menggabungkan ciri-ciri kerana kemampuannya menilai ciri-ciri tanpa mengambil kira ia dari domain yang mana. Fungsi kecergasan *Wrapper* GA diubah suai dengan menambah satu fungsi yang berperanan untuk mengekalkan keseimbangan dalam ciri-ciri yang dipilih. Nilai penalti dikenakan apabila terdapat ketidakseimbangan dalam ciri-ciri yang dipilih. Oleh itu, prestasi pengkelasan sistem berdasarkan fungsi kecergasan yang bergantung kepada peratusan sampel dikelaskan dengan betul dan keseimbangan ciri-ciri. Beberapa teknik dan data penanda aras digunakan untuk mengesahkan keberkesanan kaedah yang dicadangkan berbanding dengan sistem unimodal dan kaedah pemilihan ciri-ciri yang biasa. Hasil kajian ini telah menunjukkan bahawa prestasi pengkelasan menggunakan kaedah dicadangkan lebih optimum dengan ketepatan tertinggi 97.50%. Di samping itu, kepentingan kedua-dua biometrik sewaktu proses pengkelasan dapat dikekalkan dengan cara mengekalkan keseimbangan ciri-ciri yang dipilih.

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