

OPTIMIZATION OF LBP PARAMETERS

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Motivation

- ▶ Comparison of three types of Local Binary Patterns (LBP) as the texture features for face recognition
 - LBP, LGP, NRLBP
- ▶ Biometric system parameters optimization
 - Genetic algorithm
- ▶ Evaluation of experimental results
 - Dimension reduction
 - Recognition accuracy

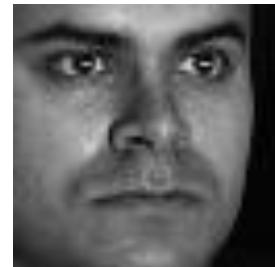
Input Data

- ▶ CMU PIE face database
 - Collected at Carnegie Mellon University

- Pose, illumination, expression

- 68 individuals
- 64x64 pixels
- 97 images per subject

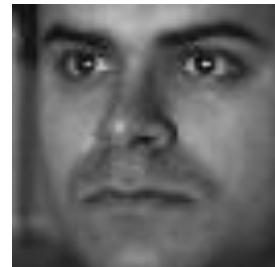
C05



C27



C29



- $68 \times 6 = 408$ training samples
- $68 \times 91 = 6188$ testing samples



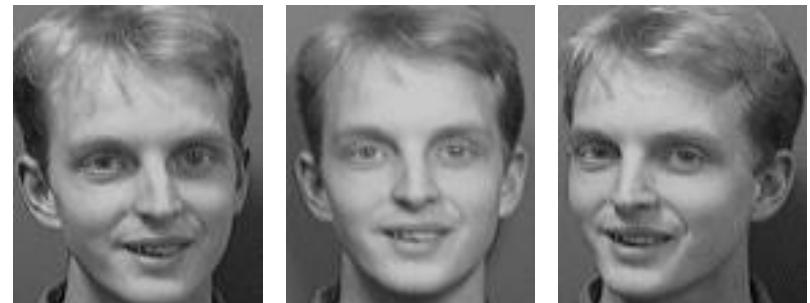
Input Data

- ▶ Cropped Yale face database
 - Collected at University of California, San Diego
 - Illumination (different positions of illumination source)
 - 38 individuals
 - 192x168 pixels
 - 68 images per subject
 - $38 \times 6 = 228$ training samples
 - $38 \times 62 = 2356$ testing samples



Input Data

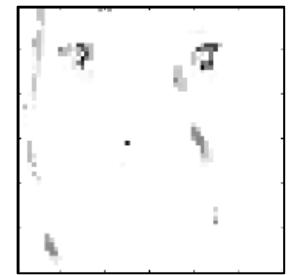
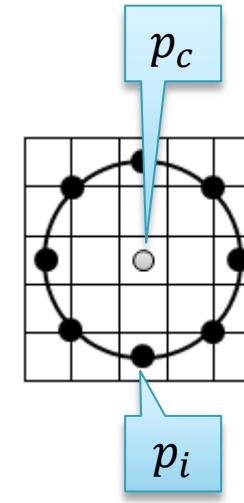
- ▶ ORL or ATT face database
 - Collected Cambridge University Computer Laboratory
 - Pose
 - 40 individuals
 - 112x92 pixels
 - 10 images per subject
 - $40 \times 6 = 240$ training images
 - $40 \times 4 = 160$ testing images



Feature extraction

▶ LBP – Local Binary Patterns

$$LBP_{P,R} = \sum_{i=0}^{P-1} s(p_i - p_c) 2^i$$
$$s(x) = \begin{cases} 1, & x \geq 0 \\ 0, & x < 0 \end{cases}$$



▶ LGP – Local Gradient Patterns

$$LGP_{P,R} = \sum_{i=0}^{P-1} s(g_i - \bar{g}) 2^i$$

$$g_i = |p_i - p_c| \quad \bar{g} = \frac{1}{P} \sum_{i=0}^{P-1} g_i$$

▶ NRLBP – Non-redundant LBP

$$NRLBP = \min(LBP_{P,R}, 2^P - 1 - LBP_{P,R})$$



The number of histogram bins (P=8, R=2)

Type	Patterns			
	Common (none)	Uniform (U2)	Rotation invariant (RI)	Rotation invariant Uniform (RIU2)
LBP	256	59	36	10
LGP	256	59	36	10
NRLBP	128	30	35	9

Feature classification

- ▶ L1 – Manhattan or city block distance

$$L1(x, y) = \sum_{i=0}^{N-1} |x_i - y_i|$$

- ▶ L2 – Euclidian distance

$$L2(x, y) = \sqrt{\sum_{i=0}^{N-1} (x_i - y_i)^2}$$

- ▶ X2 – Chi-square distance

$$\chi^2 = \sum_{i=0}^{N-1} \frac{(x_i - m_i)^2}{m_i} \quad m_i = \frac{x_i - y_i}{2}$$

- ▶ EMD – Earth mover's distance for histograms

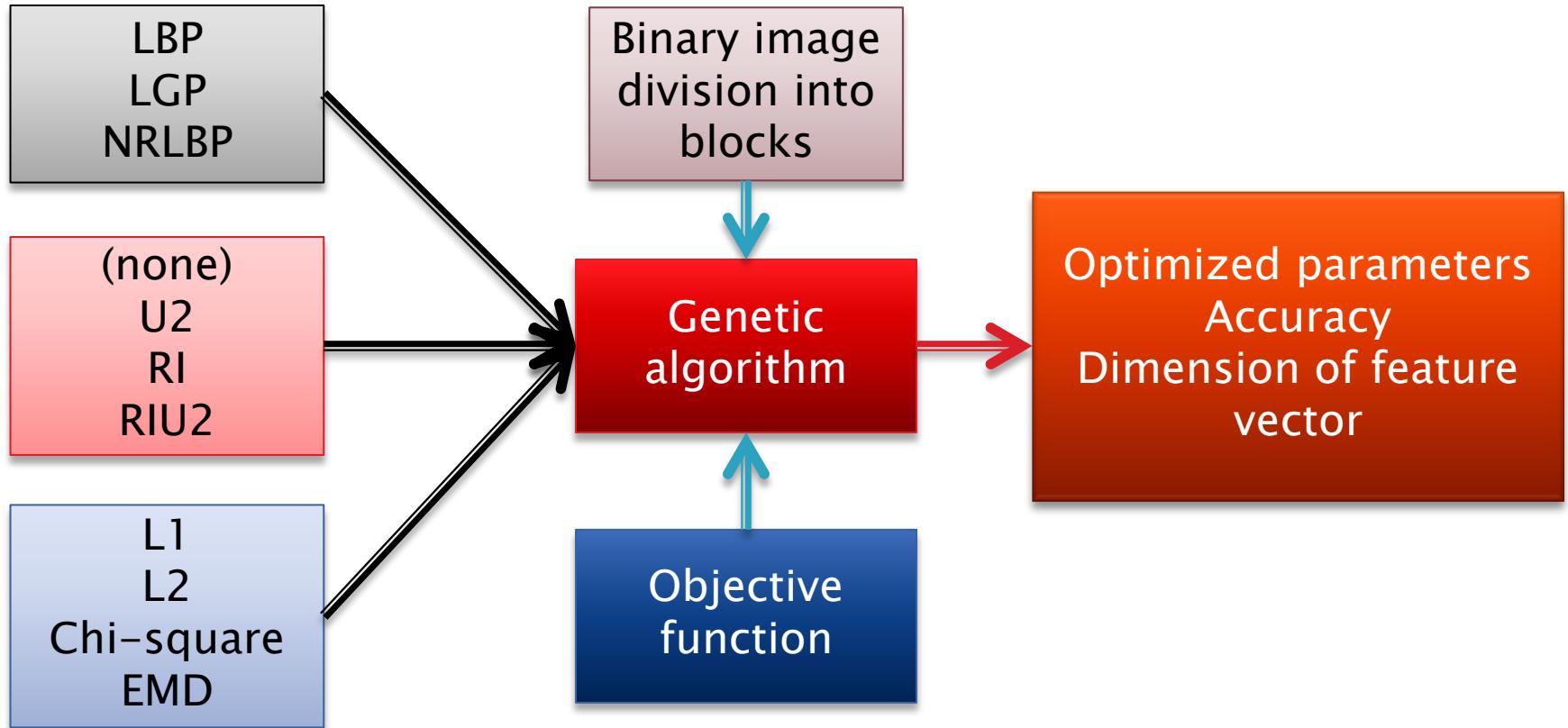
$$EMD = \min \sum_{ij} g_{ij} d_{ij}$$

$$\sum_i g_{ik} - \sum_j g_{kj} = |x(k) - y(k)|$$

Genetic algorithm

- ▶ Chromosome (sequence of values which will be optimized)
- ▶ Number of individuals in population per generation ($ps=20$)
- ▶ Number of generations ($ng=40$)
- ▶ Mutation probability ($mp=0.8$)
- ▶ Recombination probability (crossing-over) ($rp=0.5$)
- ▶ Definition of objective function – fitness
 - $f(d, x, y, acc) = 0.25 \frac{d}{xy} + (1 - acc)$

Proposed methodology



Experimental Results (CMU PIE)

Image size	Feature type	Number of blocks; Size of one block	Feature dimension	Distance measure	Recognition accuracy [%]
16x16	LBP RIU2	6 blocks; 2*12pix	60	χ^2	50.323
24x24	LBP RIU2	20 blocks; 1*20pix	200	χ^2	76.131
32x32	LBP RIU2	28 blocks; 1*28pix	280	χ^2	82.708
48x48	LBP RIU2	44 blocks; 1*44pix	440	L1	85.326
64x64	LBP RIU2	60 blocks; 1*60pix	600	χ^2	82.902

Experimental Results (YALE)

Image size	Feature type	Number of blocks; Size of one block	Feature dimension	Distance measure	Recognition accuracy [%]
24x21	LBP RIU2	20 blocks 1*17pix	200	χ^2	80.127
48x42	LBP RIU2	44 blocks 1*38pix	440	χ^2	94.056
72x63	LBP RIU2	34 blocks 2*59pix	340	χ^2	93.013
96x84	LBP RIU2	46 blocks 2*80pix	460	χ^2	91.062

Experimental Results (ORL)

Image size	Feature type	Number of blocks; Size of one block	Feature dimension	Distance measure	Recognition accuracy [%]
28x23	NRLBP U2	3 blocks 8*19pix	90	L2	99.375
42x35	LBP U2	1 block 38*31pix	59	L1	99.375
56x46	LBP U2	2 blocks 26*42pix	118	χ^2	100.000
84x69	NRLBP RIU2	20 blocks 20*13pix	180	L1, χ^2	100.000
112x92	NRLBP RIU2	9 blocks 12*88pix	81	χ^2	100.000

Conclusion

- ▶ Oft-repeated LBP using RIU2 mapping
- ▶ Consider NRLBP using RIU2 mapping
- ▶ Binary image divided along rows
- ▶ Oft-repeated Chi-square distance
- ▶ Input image 50*50 pixels ($P=8$, $R=2$)
- ▶ EMD is not suitable distance measure
- ▶ LGP is not discriminative enough
- ▶ LGP should be combined with histogram of oriented gradients (HOG)

Thank you for your attention!