

LBP and GLCM

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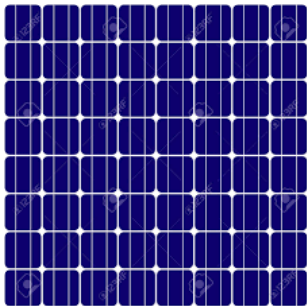
1 Texture Descriptors

- Local Binary Pattern
- Gray Level Co-occurrence Matrix(GLCM)

- ▶ Texture: **Near** periodical arrangement of pixel values or patches in image
- ▶ Textures are important discriminating features



- ▶ Textures of leaves are different from that of leaves with some disease
- ▶ Textures of solar panel with defects are different from without defect

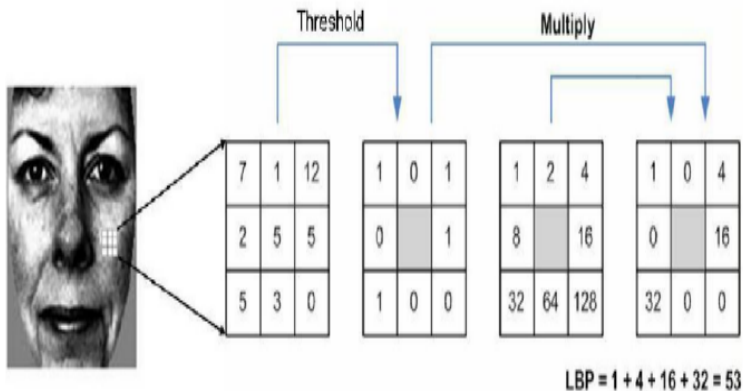




Local Binary Pattern for a pixel (x, y) in an image f

- ▶ Consider 3×3 window centered at (x, y) in f , say w
- ▶ Define:
 $w'(x, y) = 1$ if $w(x, y) > \text{pixel value at the center of } w$
 $w'(x, y) = 0$ otherwise
- ▶ Arrange the bits from TopLeft, moving clockwise
- ▶ Convert the binary sequence to decimal, which is called as LBP for the pixel (x, y)

Local Binary Pattern (cont.)





How to find LBP for an image or a segment of an image

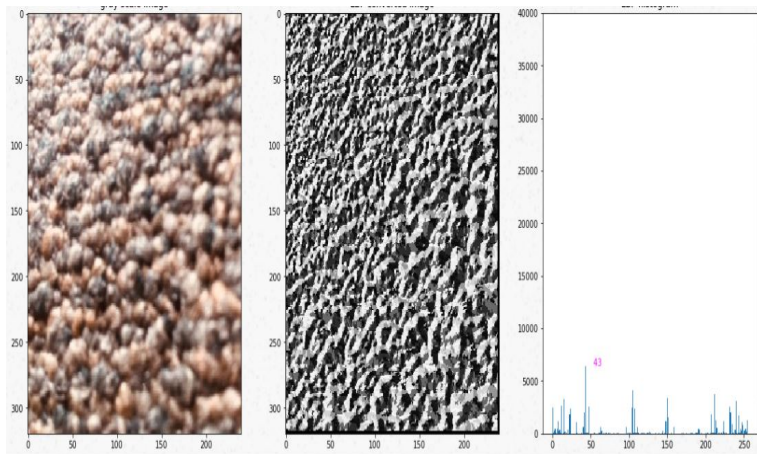
Method 1:

- ▶ Find LBP for each pixel (x, y)
- ▶ Find the histogram of all those LBP values

Issues with the method 1:

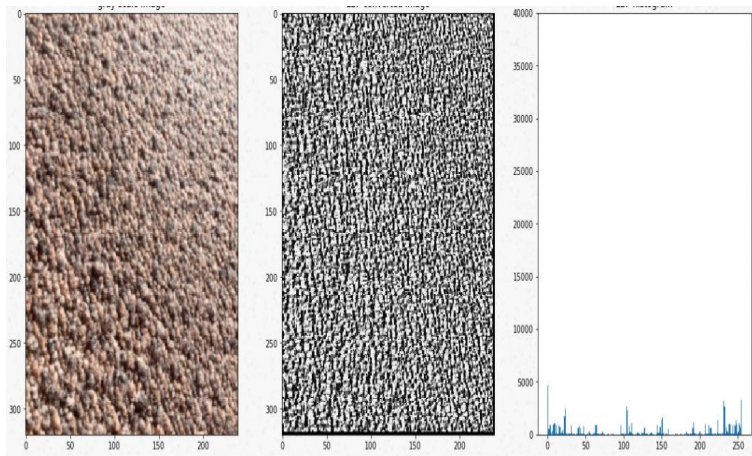
- ▶ The local pattern is not captured effectively
 - As LBP of pixels in different locations are aggregated

Local Binary Pattern (cont.)



Image; LBP; Histogram

Local Binary Pattern (cont.)



Image; LBP; Histogram



Method 2:

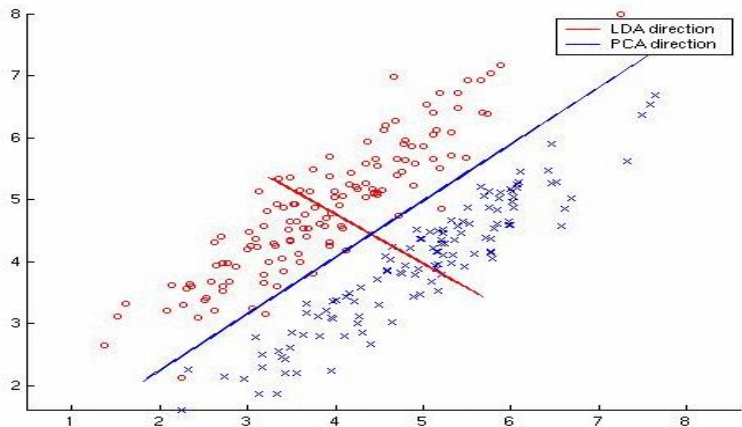
i/p: Image/Segment of image o/p: feature vector fv that describes image/segment

- ▶ Divide the image into blocks of size 16×16
- ▶ For each block B_i $1 \leq i \leq k$
 - Find LBP for each pixel $(x, y) \in B_i$
 - Find the histogram of all those LBP values, say h_i
 - Normalize h_i , say the resultant is H_i
- ▶ Concatenate all H_i s , and output $fv = (H_1, H_2, \dots H_k)$



- ▶ Size of the feature vector $256K$, where k is number of blocks
- ▶ Is there a way to reduce the size of the feature vector, and also to increase the discrimination capability
 - PCA : Transform n dimensional feature vectors to k dimension such that variance of the transformed vectors is maximum
 - LDA: Transforms n dimensional feature vector to k dimensional vector such that
 - 1) inter class variance is high
 - 2) intra class variance is low
 - ICA: Transforms n dimensional feature vectors to k dimensional vectors such that components(features) of the reduced feature vector are statistically independent

PCA vs LDA





Some Applications of LBP features

- ▶ To Classify the objects, when different objects have different texture
- ▶ SIFT descriptor for each key point may be computed as LBP of the patch centered at the point

Gray Level Co-occurrence Matrix(GLCM)



What is GLCM:

- ▶ i/p: Gray scale image;
- ▶ o/p: Gray scale image representing frequency or probability of co-occurrence of pixel values i and j as neighbours in a fixed direction d

When the direction is diagonal

2	1	2	0	1
0	2	1	1	2
0	1	2	2	0
1	2	2	0	1
2	0	1	0	1



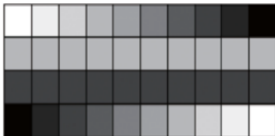
$$P_d = \begin{array}{ccc|c} 0 & 2 & 2 & 0 \\ 2 & 1 & 2 & 1 \\ 2 & 3 & 2 & 2 \\ \hline & 0 & 1 & 2 \\ & & j & \end{array} \quad \begin{array}{c} \\ i \\ \end{array}$$

Gray Level Co-occurrence Matrix(GLCM) (cont.)



When the direction is horizontal

(a) Input Image



(b) Gray Levels of Input Image

10	9	8	7	6	5	4	3	2	1
8	8	8	8	8	8	8	3	8	8
4	4	4	4	4	4	4	4	4	4
1	2	3	4	5	6	7	3	9	10

Reference Pixel

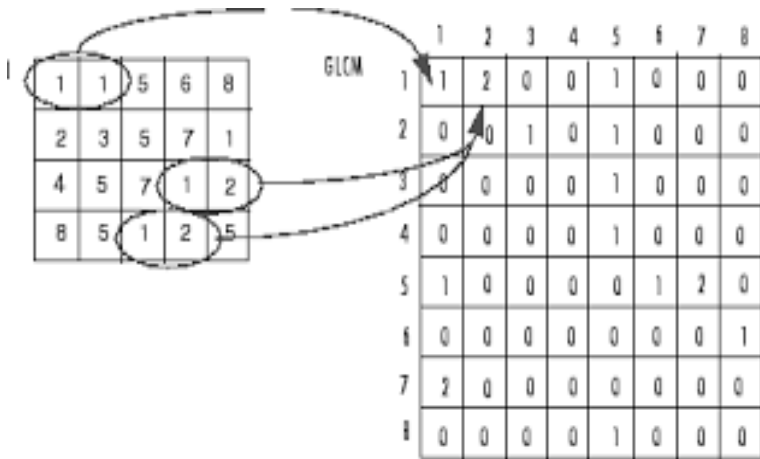
Neighbor Pixel

(c) Gray Level Co-Occurrence Matrix (GLCM)

	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	0	0	0	0	0	0	0
3	0	1	0	1	0	0	0	0	0	0
4	0	0	1	4	1	0	0	0	0	0
5	0	0	0	1	0	1	0	0	0	0
6	0	0	0	0	1	0	1	0	0	0
7	0	0	0	0	0	1	0	1	0	0
8	0	0	0	0	0	0	1	9	1	0
9	0	0	0	0	0	0	0	1	0	1
10	0	0	0	0	0	0	0	0	1	0

$$P_{ij} = P_{8,9} = 1$$

Gray Level Co-occurrence Matrix(GLCM) (cont.)

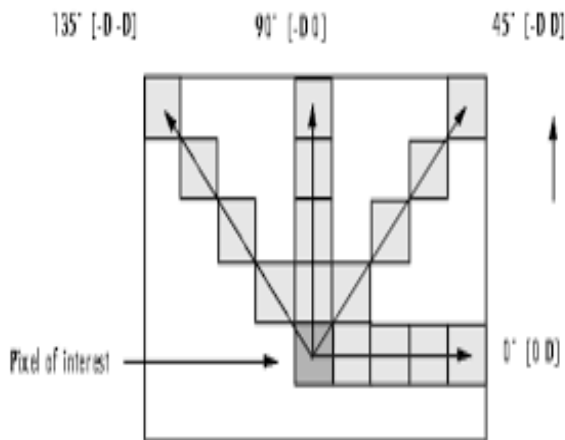




How many directions are to be considered

- ▶ Any number of directions can be considered
- ▶ For each direction, one GLCM matrix will be obtained
- ▶ Set of directions need to be chosen such that Statistical metrics on GLCM in such directions are
 - Very different for different objects
 - Similar for similar objects

Gray Level Co-occurrence Matrix(GLCM) (cont.)





Some Statistical measures based on GLCM

Let $P = (p_{i,j})$ be the GLCM, where $p_{i,j}$ is the probability of co-occurrence of i and j in a particular direction

- ▶ Energy: $\sum p_{i,j}^2$
 - Measures the energy of co-occurrence
 - Energy is more for some distributions, and less for some distributions
- ▶ Entropy: $-\sum p_{i,j} \ln(p_{i,j})$
 - Measures the uncertainty of co-occurrence of i and j
- ▶ Contrast: $\sum p_{i,j} (i - j)^2$
 - Measures the contrast between i and j
- ▶ Homogeneity: $\sum \frac{p_{i,j}}{1 + |i - j|}$



- Measures the closeness of the elements in the GLCM to GLCM diagonal
- ▶ Correlation: $\sum \frac{(i-\mu_i)(j-\mu_j)}{\sigma_i\sigma_j} p_{i,j}$
 - Measures the correlation between i and j , when the co-occur
- ▶ Mean of i , when i and j co-occur: $\sum_{i,j} i p_{i,j}$
- ▶ Variance of i , when i and j co-occur: $\sum_{i,j} (i - \mu_i)^2 p_{i,j}$
- ▶ Kth moment of element difference: $\sum (i - j)^k p_{i,j}$

Gray Level Co-occurrence Matrix(GLCM) (cont.)



How to find contrast

0	1	4	9
1	0	1	4
4	1	0	1
9	4	1	0

Contrast weight

*

0.167	0.083	0.042	0
0.083	0.167	0	0
0.042	0	0.25	0.042
0	0	0.042	0.083

Horizontal GLCM

=

0	0.083	0.168	0
0.083	0	0	0
0.168	0	0	0.042
0	0	0.042	0



$$\begin{aligned} &= .083 + .168 + .083 + .168 + .042 + .042 \\ &= .586 \end{aligned}$$

Gray Level Co-occurrence Matrix(GLCM) (cont.)



How to find homogeneity

1	0.5	0.333	0.25
0.5	1	0.5	0.333
0.333	0.5	1	0.5
0.25	0.333	0.5	1

similarity weight

*

0.167	0.083	0.042	0
0.083	0.167	0	0
0.042	0	0.25	0.042
0	0	0.042	0.083

Horizontal GLCM

=

0.167	0.042	0.014	0
0.042	0.167	0	0
0.014	0	0.25	0.021
0	0	0.021	0.083



$$\begin{aligned} &0.167+0.042+0.014+0.042+0.167+ \\ &0.014+0.25+0.021+0.021+0.083 \\ &= 0.804 \end{aligned}$$

Graphical view of GLCM of various textures

