Digital Image Processing

Lecture #5

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Announcement

- Class Information
  - Homework #2
    - Due at **11:59 am on Apr. 12, 2016**

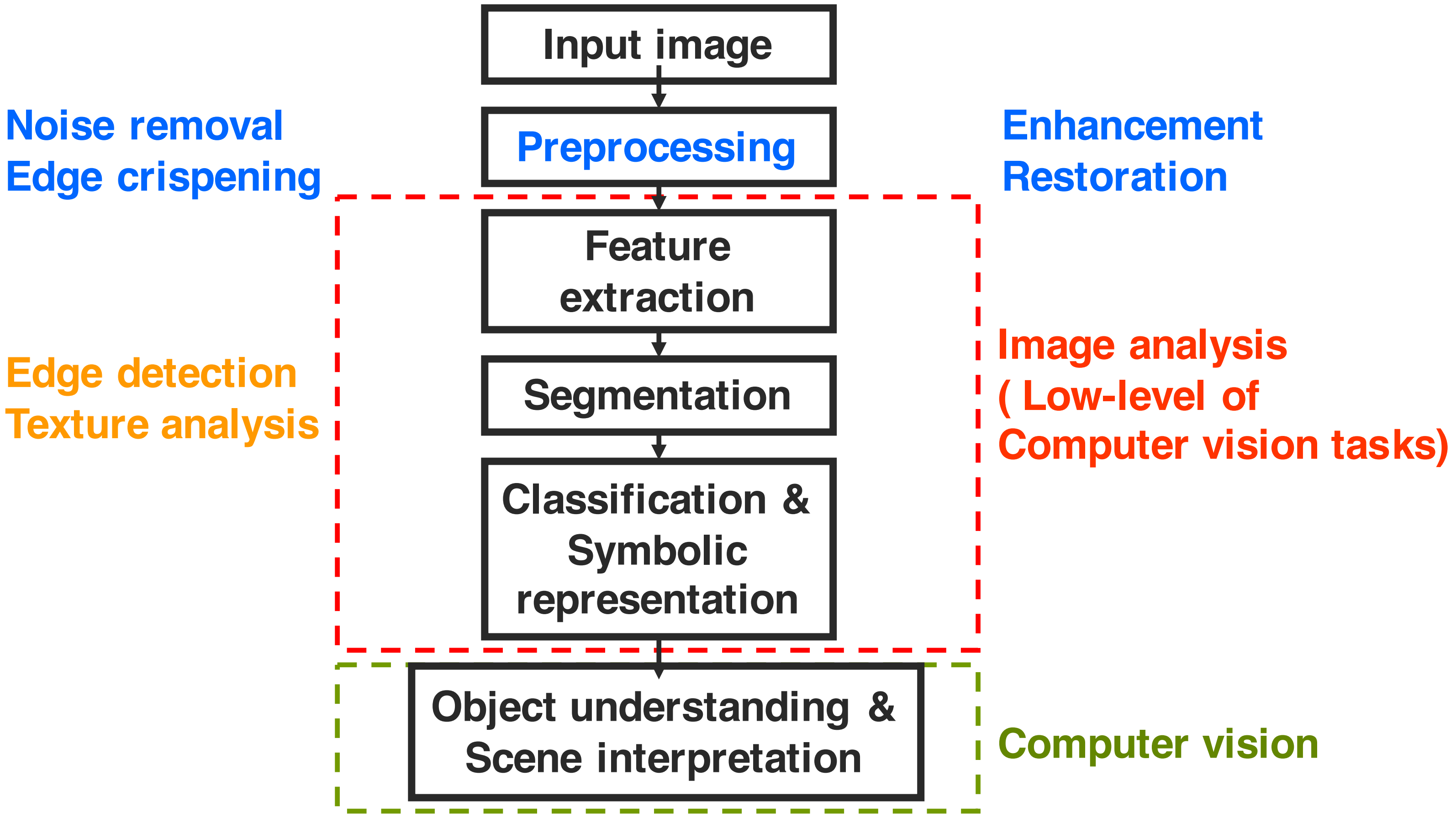
Announcement

- Seminar Talk
  - “Recreational Graphics: Applications in Life, Art, and Entertainment”
    - 03/25 02:20~03:30 p.m. @ 德田館 R103
    - Prof. Hung-Kuo Chu (James)
    - Department of Computer Science,
      National Tsing Hua University

Texture Analysis

Texture Analysis

- Image analysis and its applications



Texture Analysis

- What is texture?

Texture Analysis

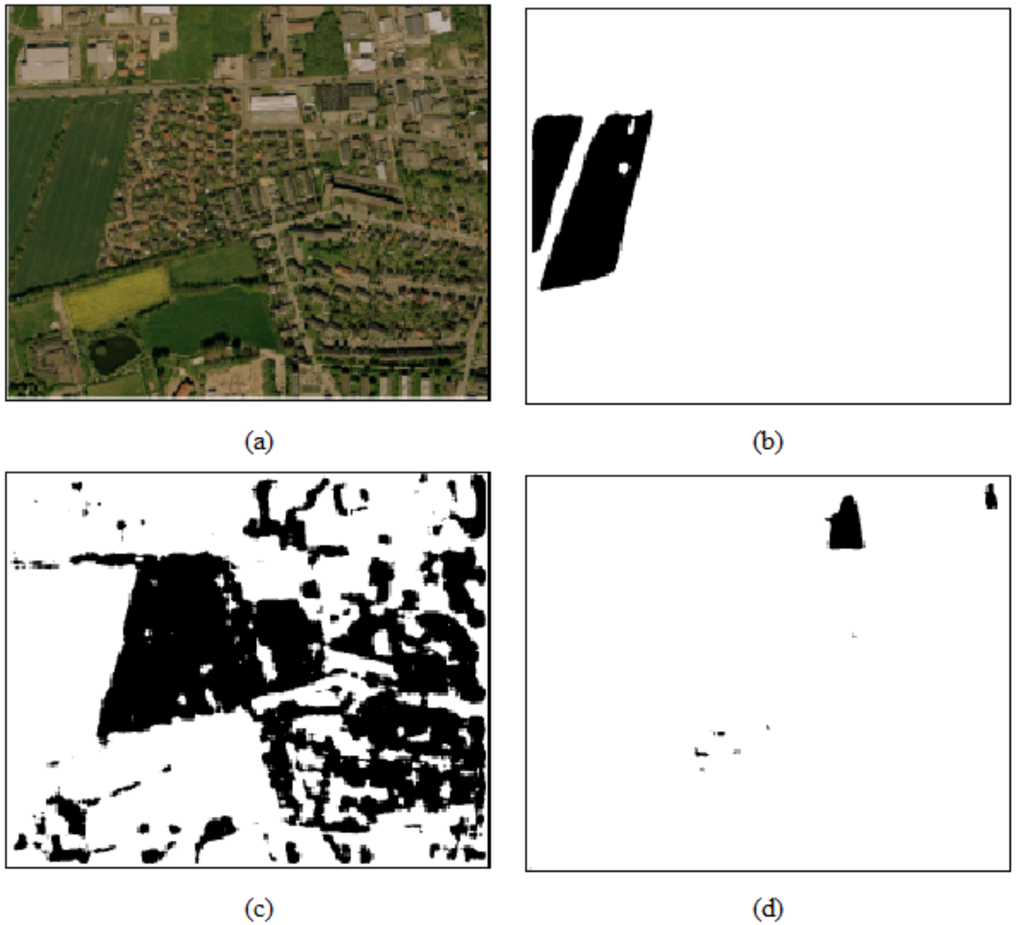
- What is texture?
  - No mathematical definition
  - Two dimensional arrays of variations
  - Semi-regular structured patterns of object
  - E.g. Surfaces such as sand, grass, wool, cloth, leaves, etc.

Texture Analysis

- Why texture analysis?
  - People started to be interested in late 50's and early 60's
    - Analyze aerial images / texture patches

Texture Analysis

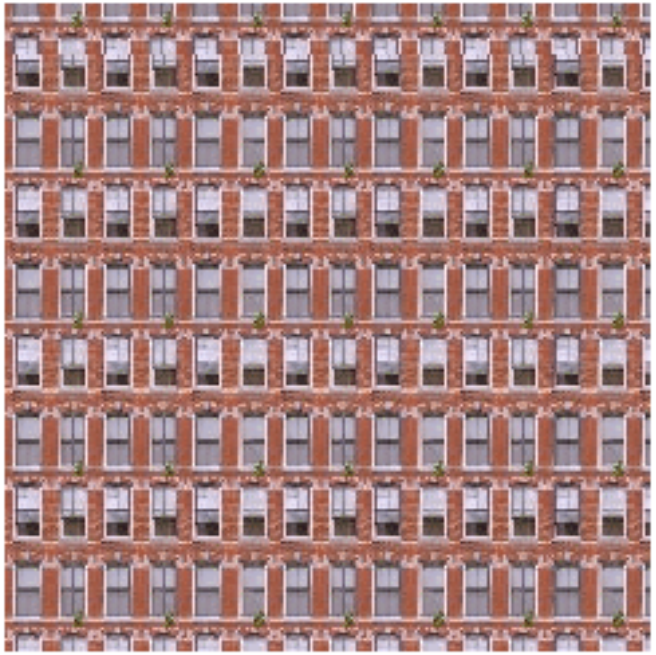
- Example (an aerial image)

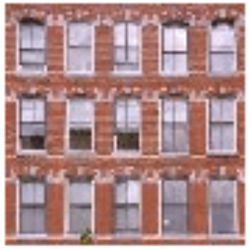


(a) Aerial photo (b)Field (c) Residential area (d) Vegetation area

Texture Analysis

- Example (Texture Synthesis)





Texture Analysis

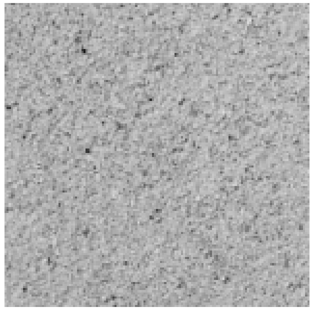
- History of texture analysis
  - Fourier Spectral Methods
  - Edge Detection Methods
  - Autocorrelation Methods
  - Decorrelation Methods
  - Dependency Matrix Method

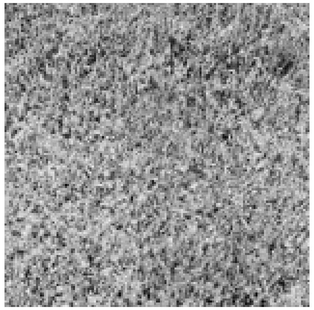
Texture Analysis

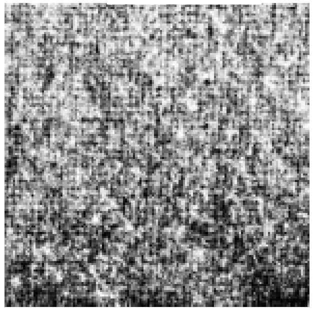
- Fourier Spectral Methods
  - Right direction but incomplete development
  - No continuous work for a long while
- Edge Detection Methods
  - Edge detection
  - Use edge density and orientation as texture features

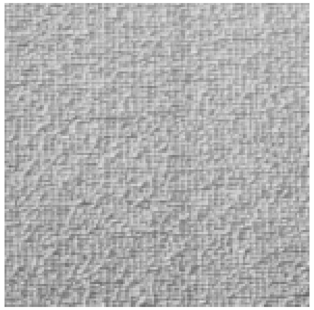
Texture Analysis

- **Autocorrelation Methods**
  - Treat the texture pattern as a 2D random process, denoted as F(x,y)
  - Statistical approach


Sand


Grass

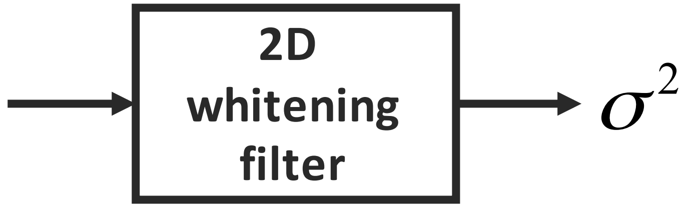

Wool


Raffia

$$E\{F(x,y)F(x-\Delta x, y-\Delta y)\}$$

Texture Analysis

- Decorrelation Methods
  - 2D whitening filter
    - Special type of decorrelation operator



2D whitening filter

σ^2

$$\hat{W}(j,k) = F(j,k) \otimes H_W(j,k)$$

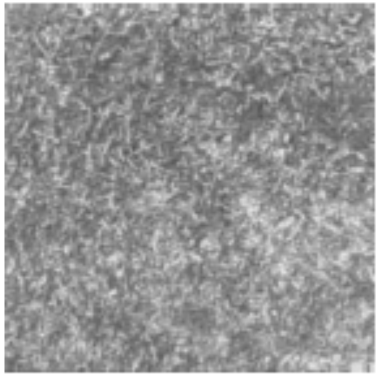
  - Spatially decorrelated
    - Form histogram as its feature

Texture Analysis

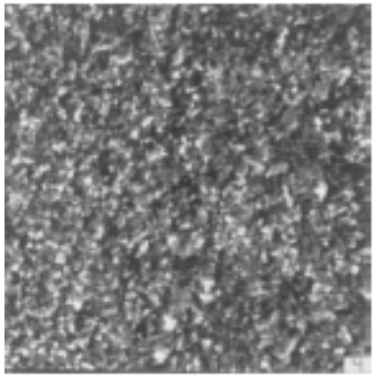
- Dependency Matrix Method
  - Joint probability
  - Also called Co-occurrence method

$$P(a,b \mid j,k,\Delta j,\Delta k)$$

$$= \text{Prob}\{F(j,k)=a, F(j-\Delta j,k-\Delta k)=b, 0 \le a,b \le L-1\}$$



Grass



Ivy

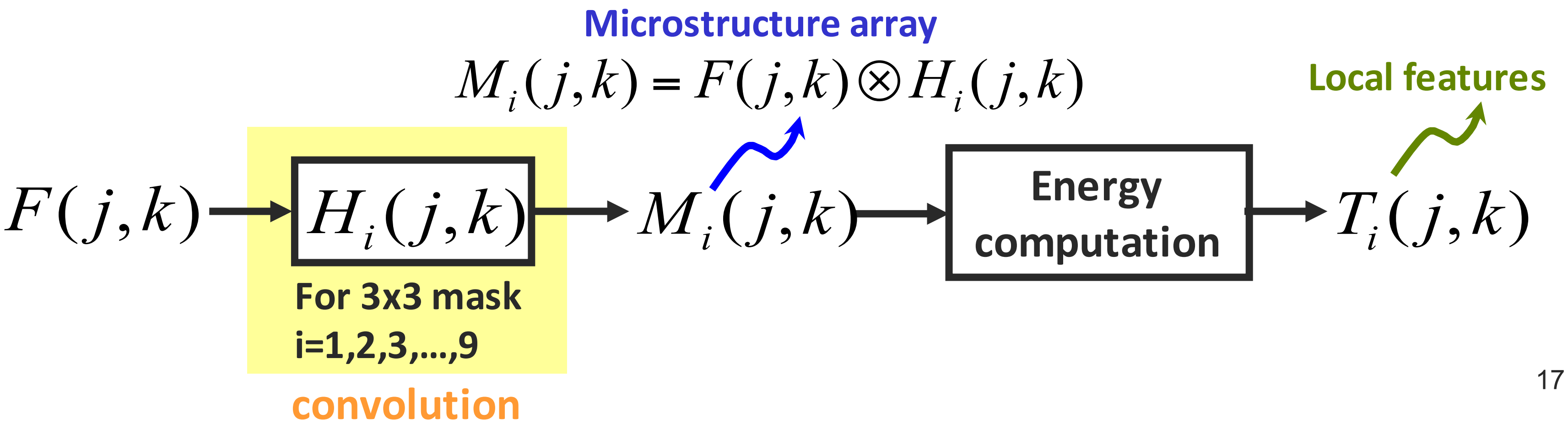
Texture Analysis

- History of texture analysis
  - Fourier Spectra methods
  - Edge Detection Methods
  - Autocorrelation Methods
  - Decorrelation Methods
  - Dependency Matrix Method

→ Not successful!!

Texture Analysis

- Laws' Method
  - Micro-structure (Multi-channel) method
    - Emphasize the microstructure of the texture
    - Two steps
      - step 1: Convolution
      - step 2: Energy computation



Microstructure array

$$M_i(j,k) = F(j,k) \otimes H_i(j,k)$$

Local features

$$F(j,k) \rightarrow \boxed{H_i(j,k)} \rightarrow M_i(j,k) \rightarrow \boxed{\text{Energy computation}} \rightarrow T_i(j,k)$$

For 3x3 mask
i=1,2,3,…,9

convolution

Texture Analysis

- **Laws' Method**
  - **//Step 1// Convolution** $M_i(j,k) = F(j,k) \otimes H_i(j,k)$
    - **Micro-structure impulse response arrays (a basis set)**

$H_i(j,k)$

for 3x3 mask, i=1,2,3,…,9

for 5x5 mask, i=1,2,3,…,25

How to choose the mask size?

$$\frac{1}{36}\begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

Laws 1

$$\frac{1}{12}\begin{bmatrix} 1 & 0 & -1 \\ 2 & 0 & -2 \\ 1 & 0 & -1 \end{bmatrix}$$

Laws 2

$$\frac{1}{12}\begin{bmatrix} -1 & 2 & -1 \\ -2 & 4 & -2 \\ -1 & 2 & -1 \end{bmatrix}$$

Laws 3

$$\frac{1}{12}\begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

Laws 4

$$\frac{1}{4}\begin{bmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{bmatrix}$$

Laws 5

$$\frac{1}{4}\begin{bmatrix} -1 & 2 & -1 \\ 0 & 0 & 0 \\ 1 & -2 & 1 \end{bmatrix}$$

Laws 6

$$\frac{1}{12}\begin{bmatrix} -1 & -2 & -1 \\ 2 & 4 & 2 \\ -1 & -2 & -1 \end{bmatrix}$$

Laws 7

$$\frac{1}{4}\begin{bmatrix} -1 & 0 & 1 \\ 2 & 0 & -2 \\ -1 & 0 & 1 \end{bmatrix}$$

Laws 8

$$\frac{1}{4}\begin{bmatrix} 1 & -2 & 1 \\ -2 & 4 & -2 \\ 1 & -2 & 1 \end{bmatrix}$$

Laws 9

Texture Analysis

- Laws' Method
  - Micro-structure impulse response arrays
    - Generated by the tensor product of the 1D horizontal and vertical masks

$$L_3 = \frac{1}{6}[1 \quad 2 \quad 1] \qquad E_3 = \frac{1}{2}[-1 \quad 0 \quad 1] \qquad S_3 = \frac{1}{2}[1 \quad -2 \quad 1]$$

Local averaging

Edge detector (1st-order gradient)

spot detector (2nd–order gradient)

- E.g.

$$L_3^T \otimes E_3 = \frac{1}{6}\begin{bmatrix}1\\2\\1\end{bmatrix} \otimes \frac{1}{2}[-1 \quad 0 \quad 1] = \frac{1}{12}\begin{bmatrix}-1 & 0 & 1\\-2 & 0 & 2\\-1 & 0 & 1\end{bmatrix}$$

Laws 2

Texture Analysis

- **Laws' Method**
  - ○ **Micro-structure impulse response arrays**
    - **1979 → 1984,1986 mathematical analysis of Laws' filters**
    - **Examine the frequency response of** $L_3,\ E_3,$ and S_3

$$L_3 = \frac{1}{6}[1 \quad 2 \quad 1]$$

Kroneckor Delta

$$h[n] = \frac{1}{6}\left(\delta[n-1] + 2\delta[n] + \delta[n+1]\right) \qquad \delta[n] = \begin{cases} 1 & n = 0 \\ 0 & otherwise \end{cases}$$

$$H(\omega) = \frac{1}{6}\left(e^{-j\omega} + 2 + e^{j\omega}\right) = \frac{2}{6}(1 + \cos\omega)$$

→ **Low-pass filter**

Texture Analysis

- Laws' Method
  - Micro-structure impulse response arrays
    - Examine the frequency response of $L_3, \ E_3, \ and \ S_3$

$$E_3 = \frac{1}{2}[-1 \quad 0 \quad 1] \qquad h[n] = \frac{1}{2}\left(-\delta[n-1] + \delta[n+1]\right)$$

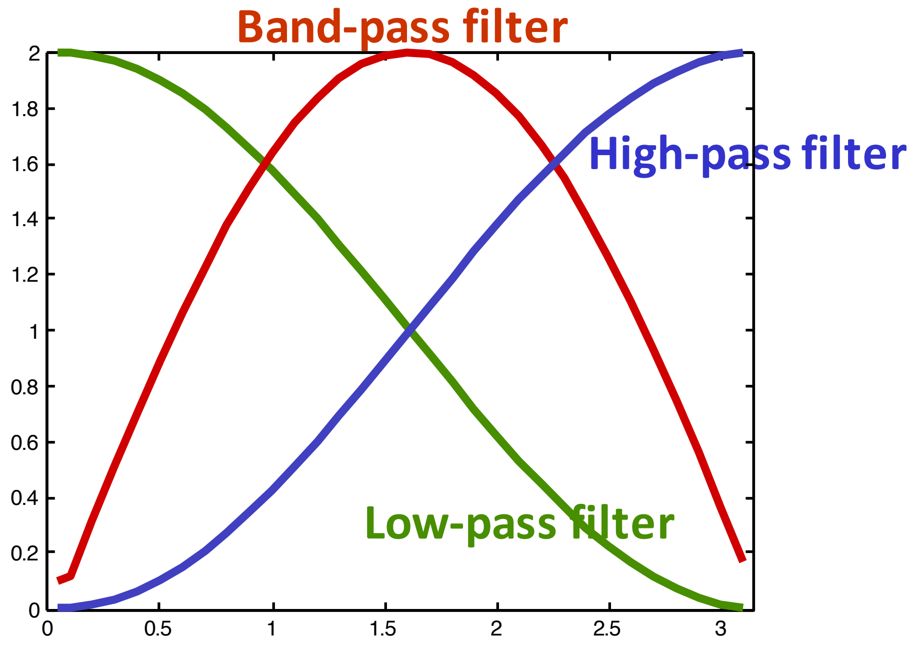
$$H(\omega) = \left(-e^{-j\omega} + e^{j\omega}\right) = 2j\sin\omega$$ → **Bandpass filter**

$$S_3 = \frac{1}{2}[1 \quad -2 \quad 1] \quad h[n] = \frac{1}{2}\left(\delta[n-1] - 2\delta[n] + \delta[n+1]\right)$$

$$H(\omega) = \frac{1}{2}\left(e^{-j\omega} - 2 + e^{j\omega}\right) = \cos\omega - 1$$ → **High-pass filter**

Texture Analysis

- Laws' Method
  - Micro-structure impulse response arrays
    - Examine the frequency response of $L_3,\ E_3,\ and\ S_3$



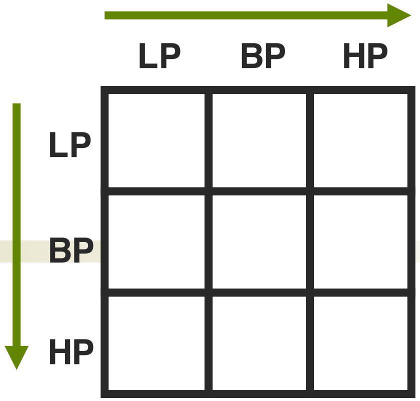
$H_i(j,k)$

for 3x3 mask, i=1,2,3,…,9

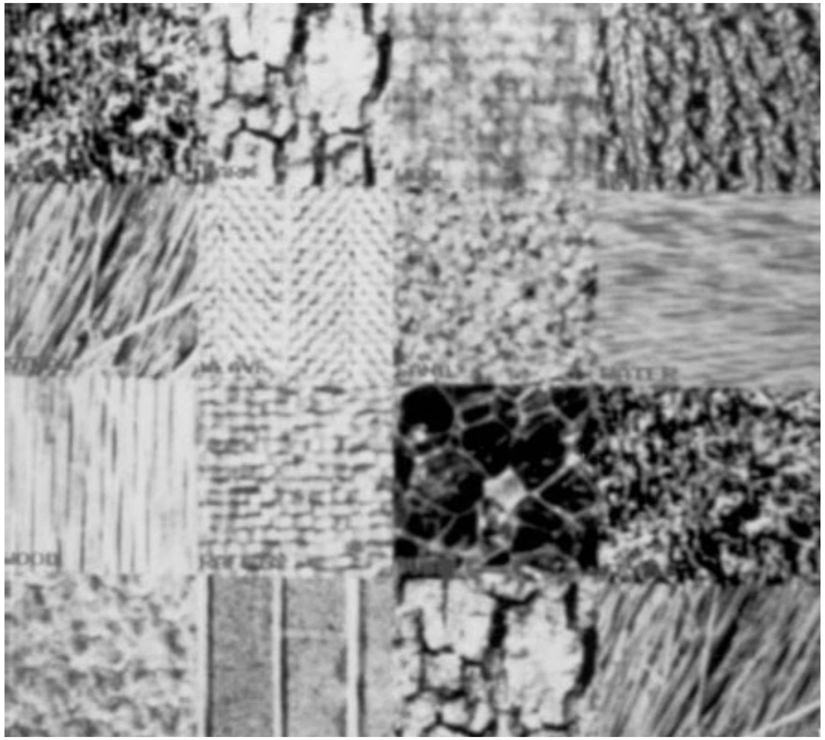
	LP	BP	HP
LP			
BP			
HP			

// Multi-channel method //

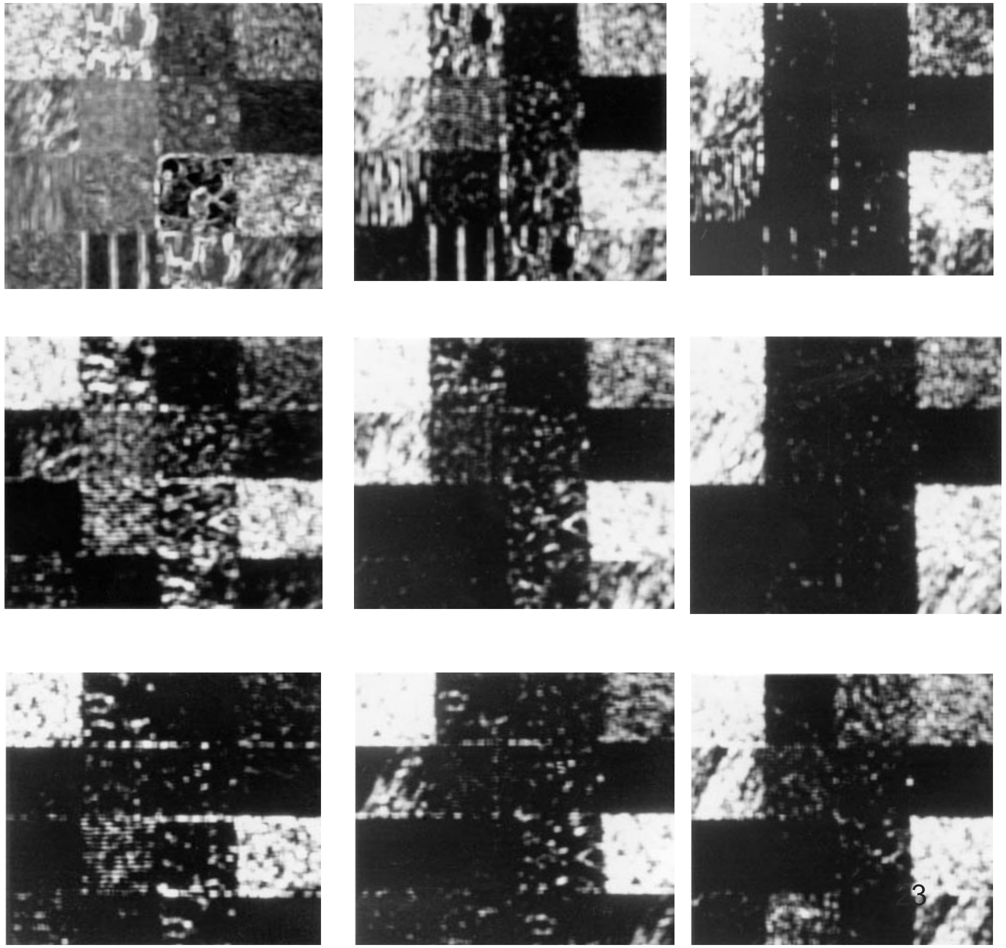
Texture Analysis



- Example



original image

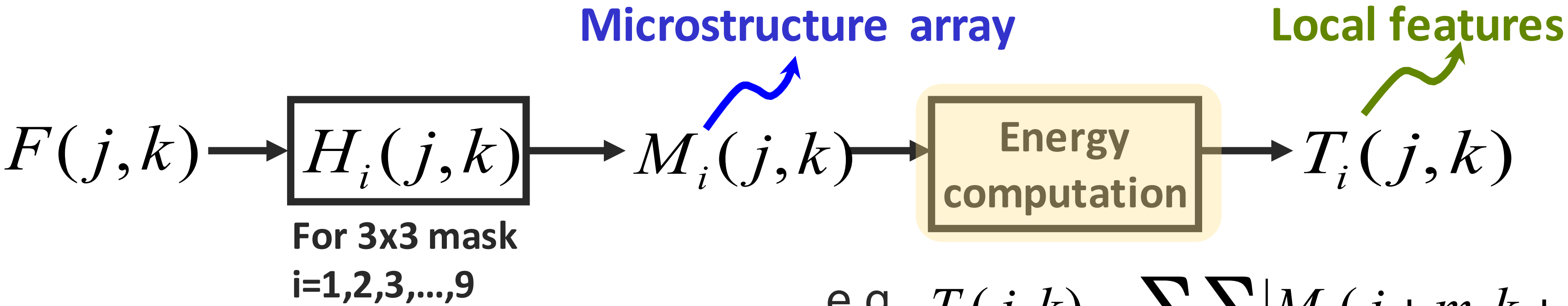


Texture Analysis

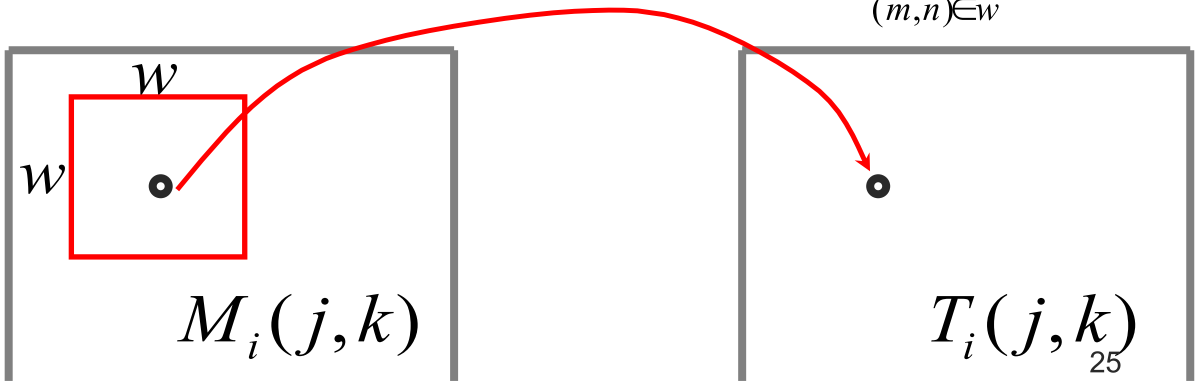
- Laws' Method
  - //Step 2// Energy Computation $T_i(j,k)$
    - Extract features over a window that has a few cycles of the repetitive texture
    - How to choose the window size?
    - Global/local energy computation
    - 9 energy features correspond to the energy in the 9 subbands. We use the energy distribution in these 9 subbands to differentiate different texture types
    - Features
      - Mean, standard deviation, energy, smoothness etc.

Texture Analysis

- **Laws' Method**
  - **//Step 2// Energy Computation**



e.g. $T_i(j,k) = \sum_{(m,n)\in w}\sum \left|M_i(j+m,k+n)\right|^2$



Texture Analysis

- Notes for Laws' method
  - How to choose the mask size? $H_i(j,k)$
  - Fixed subband structure vs Dynamic subband structure
  - How to choose the window size for energy computation?
    - For texture analysis, window size is usually set to be 13x13 or 15x15

Texture Analysis

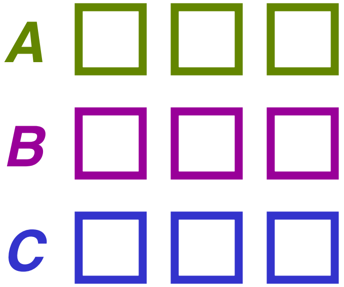
- Texture classification/segmentation
  - Given 9 feature sets, $T_1, \ T_2, \ T_3, \cdots \ T_9$

    How do we do texture classification?
  - Two cases
    - Each input is homogeneous
    - Single input consists of more than one texture
  - Two approaches
    - Supervised texture classification
    - Un-supervised texture classification

Texture Analysis

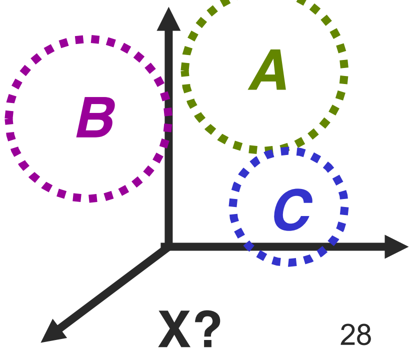
- Texture classification
  - Supervised texture classification
    - For each given texture type

$$\begin{array}{lll} textureA & \rightarrow & T_{A1},\ T_{A2},\ T_{A3},\cdots\ T_{A9} \\ textureB & \rightarrow & T_{B1},\ T_{B2},\ T_{B3},\cdots\ T_{B9} \\ textureC & \rightarrow & T_{C1},\ T_{C2},\ T_{C3},\cdots\ T_{C9} \end{array}$$



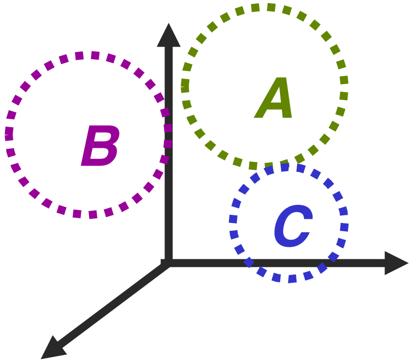
    - Texture space → 9 dimensional
    - Given texture X

      Use nearest neighbor classification rule



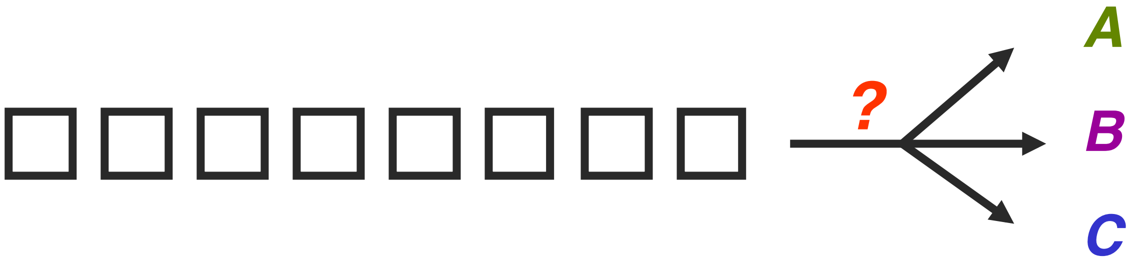
Texture Analysis

- Texture classification
  - Feature space dimension reduction
    - Not considering all 9 features equally
    - More important feature
      - More discriminating power
      - Weighted more
    - Less important feature
      - Weighted less
      - Taken out from the feature set



Texture Analysis

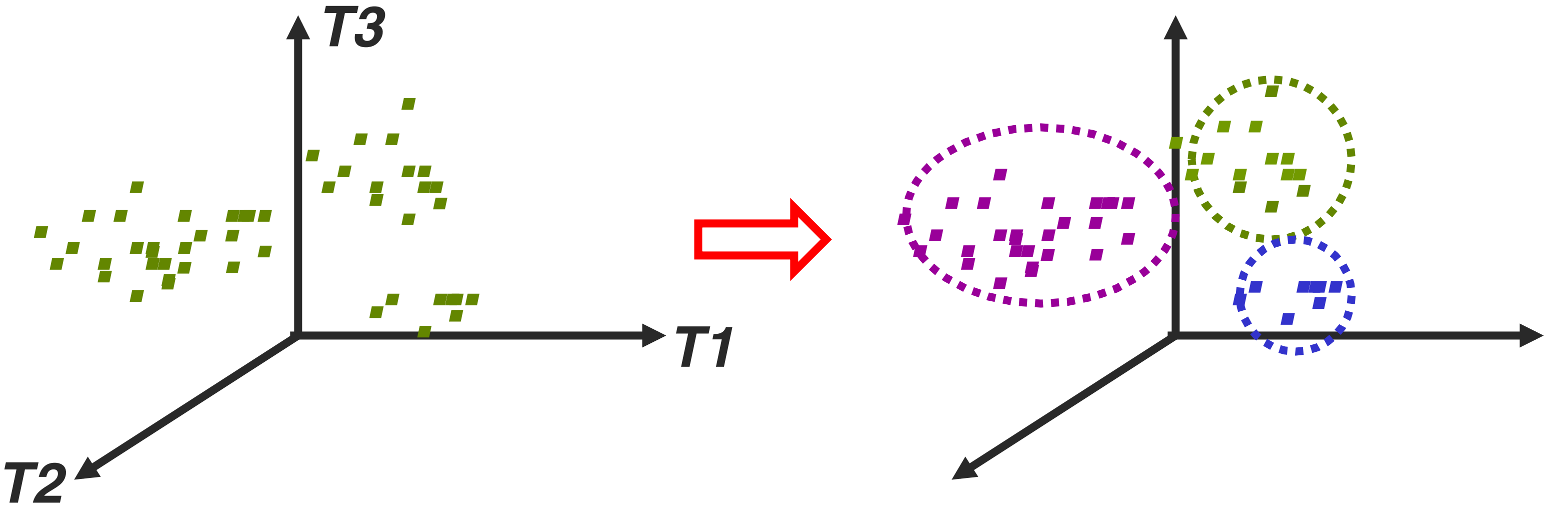
- Texture classification
  - Un-supervised texture classification
    - For several texture patches



  - K-means algorithm
    - The famous tool to handle unsupervised classification problem

Texture Analysis

- K-means algorithm
  - K=3



  - Good classification
    - Inter-clustering →
    - Intra-clustering →

Texture Analysis

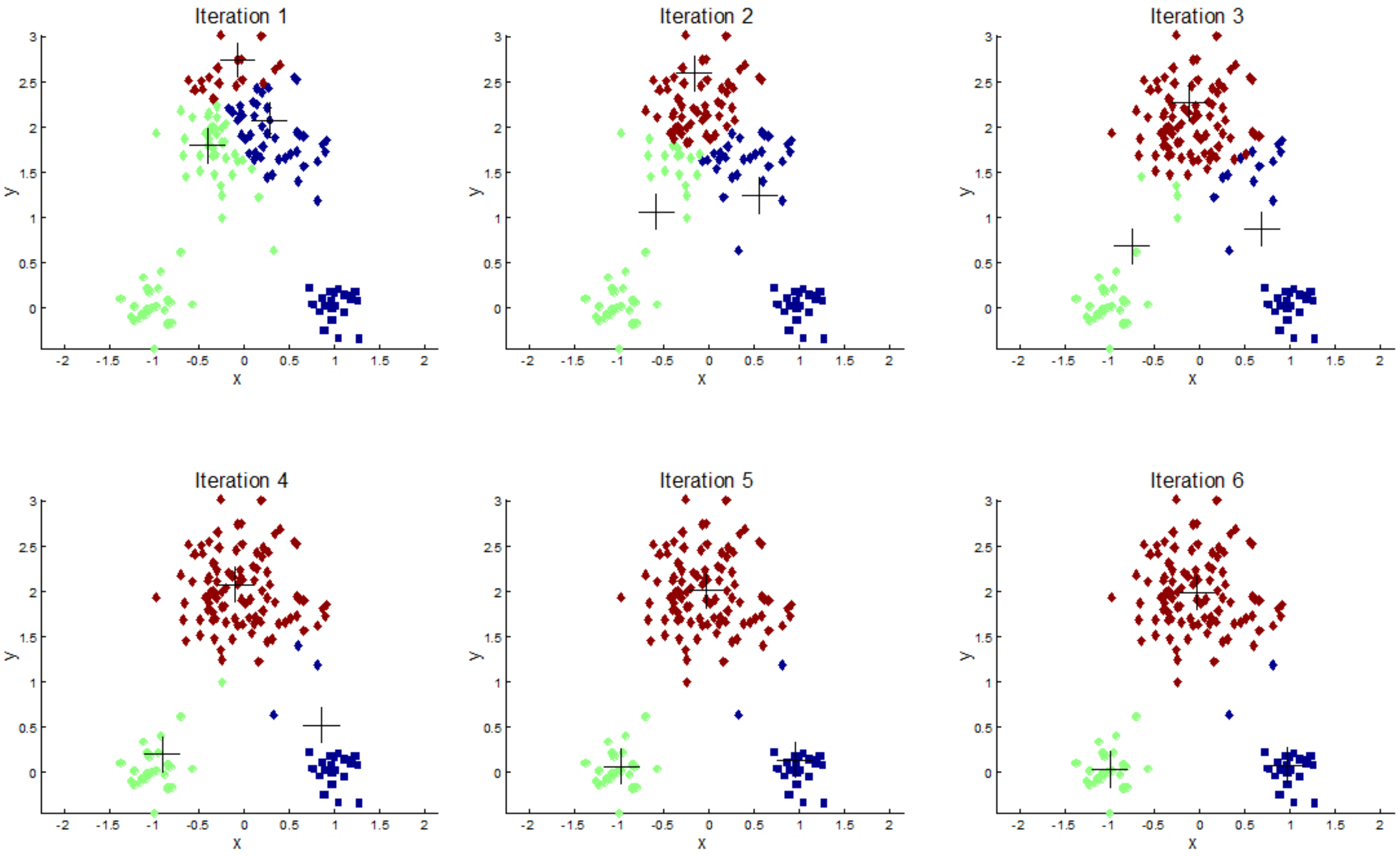
- K-means algorithm
  - Two issues
    - How to choose k?

      depends on the inter-cluster and intra-cluster statistical analysis
      OR by the problem set-up (domain knowledge)

    - Given k, how to do the clustering?
      - // Initialization //
        - Select k vectors as the initial centroids
        - Do the following iterations
      - // step1 // Form k clusters using the NN rule
      - // step2 // re-compute the centroid of each cluster

Texture Analysis

- **K-means algorithm demo**



Texture Analysis

- Texture classification
  - Two criterions
    - If pixels belong to the same type of texture, their associated feature vectors are close to each other in the feature space
    - Pixels belong to the same texture type should be close to each other in the space domain
  - What is a good segmentation result?
    - Regions of a segment should be homogeneous w.r.t. some properties (i.e. feature vectors are close to each other in the feature space)
    - Region interior should be simple and without many holes
    - Boundaries of each segment should be simple, not ragged