Object and Pattern Recognition/Machine Learning Lecture 9

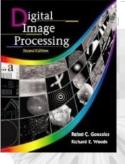
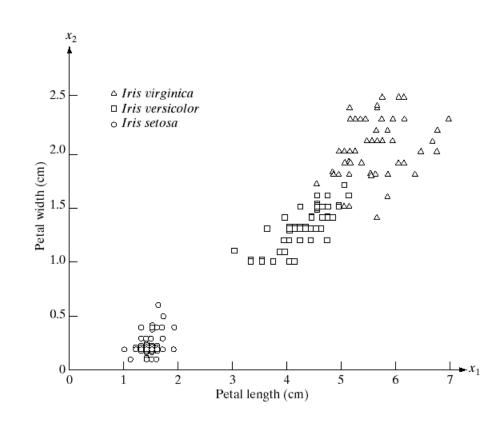
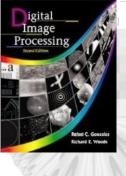
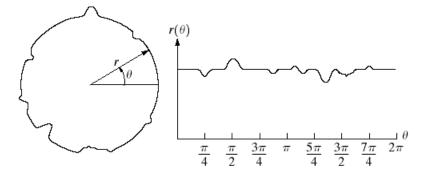


FIGURE 12.1
Three types of iris flowers described by two measurements.

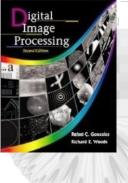


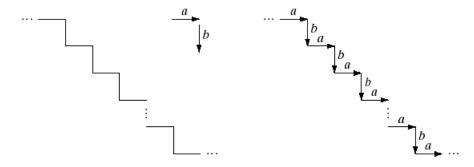




a b

FIGURE 12.2 A noisy object and its corresponding signature.





a b

FIGURE 12.3 (a) Staircase structure. (b) Structure coded in terms of the primitives a and b to yield the string description ... $ababab \dots$

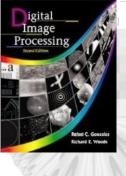
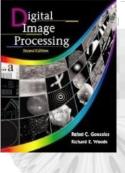




FIGURE 12.4 Satellite image of a heavily built downtown area (Washington, D.C.) and surrounding residential areas. (Courtesy of NASA.)



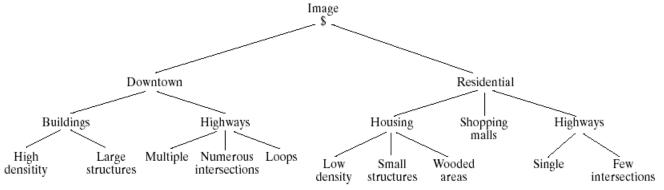
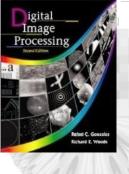


FIGURE 12.5 A tree description of the image in Fig. 12.4.



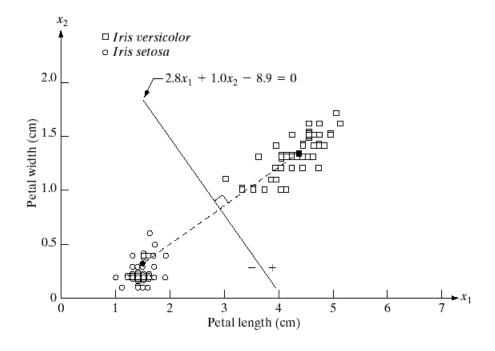
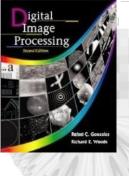


FIGURE 12.6

Decision boundary of minimum distance classifier for the classes of *Iris* versicolor and *Iris* setosa. The dark dot and square are the means.



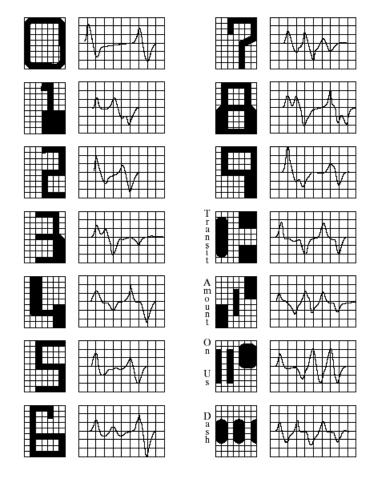
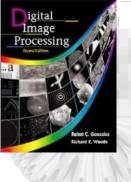


FIGURE 12.7

American Bankers Association E-13B font character set and corresponding waveforms.



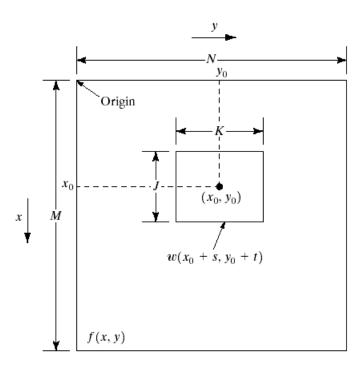
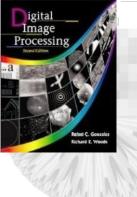
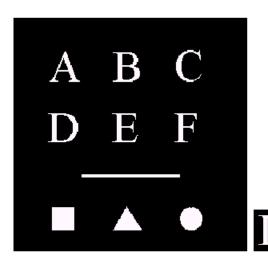
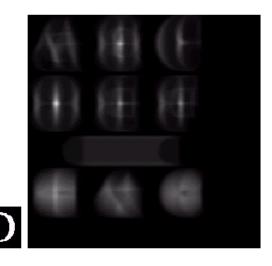


FIGURE 12.8 Arrangement for obtaining the correlation of f and w at point (x_0, y_0) .







a b c

FIGURE 12.9

(a) Image.
(b) Subimage.
(c) Correlation
coefficient of (a)
and (b). Note that
the highest
(brighter) point in
(c) occurs when
subimage (b) is
coincident with the
letter "D" in (a).

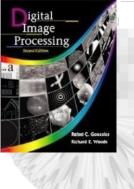
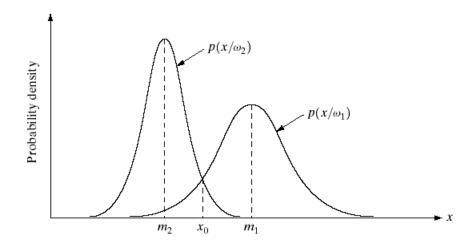
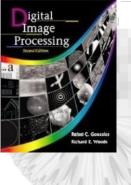


FIGURE 12.10

Probability density functions for two 1-D pattern classes. The point x_0 shown is the decision boundary if the two classes are equally likely to occur.





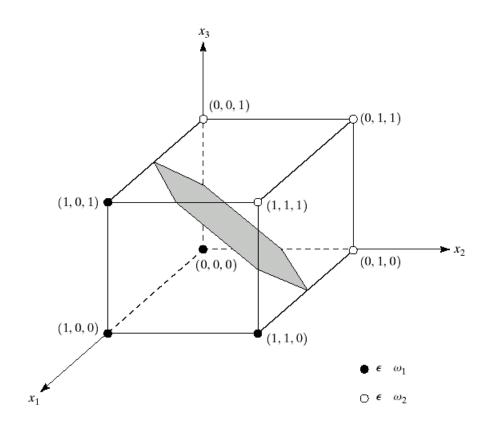


FIGURE 12.11

Two simple pattern classes and their Bayes decision boundary (shown shaded).

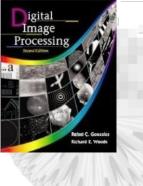
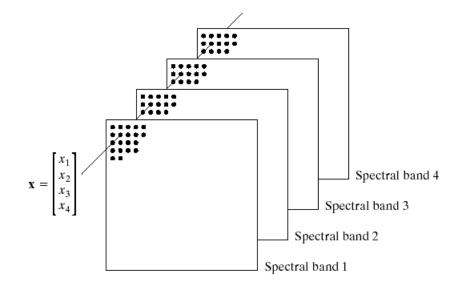
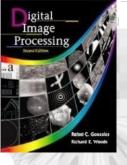


FIGURE 12.12

Formation of a pattern vector from registered pixels of four digital images generated by a multispectral scanner.





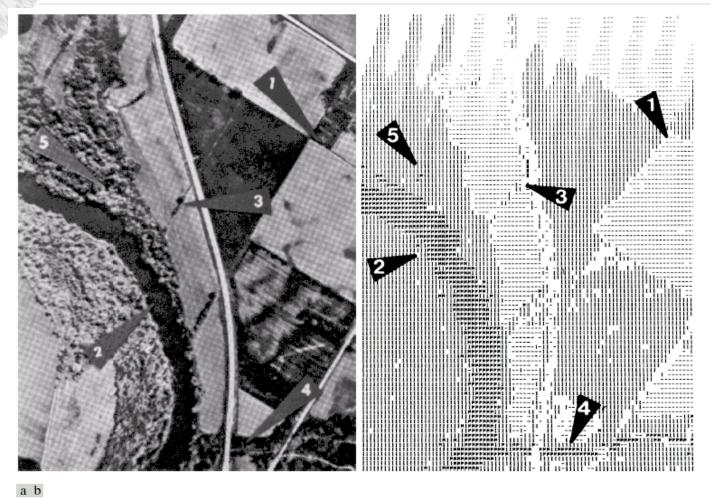
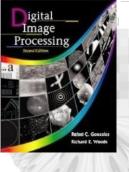
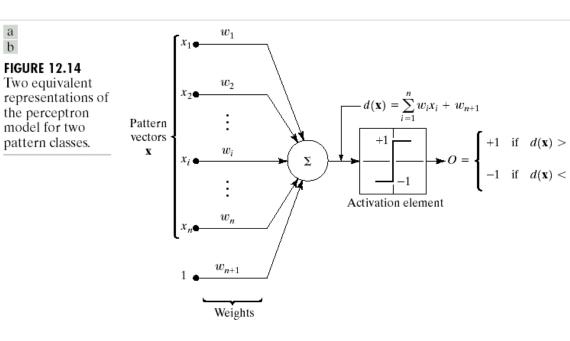
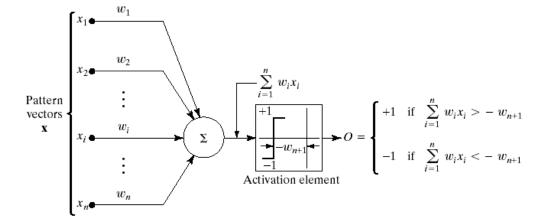


FIGURE 12.13 (a) Multispectral image. (b) Printout of machine classification results using a Bayes classifier. (Courtesy of the Laboratory for Applications of Remote Sensing, Purdue University.)

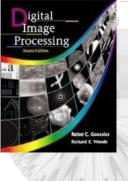




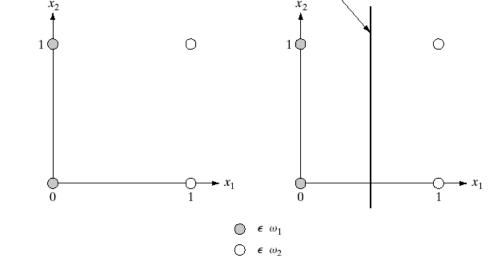




 $-d(\mathbf{x}) = -2x_1 + 1 = 0$



Object Recognition

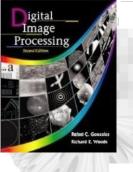


a b

FIGURE 12.15

(a) Patterns belonging to two classes. (b) Decision boundary

determined by training.



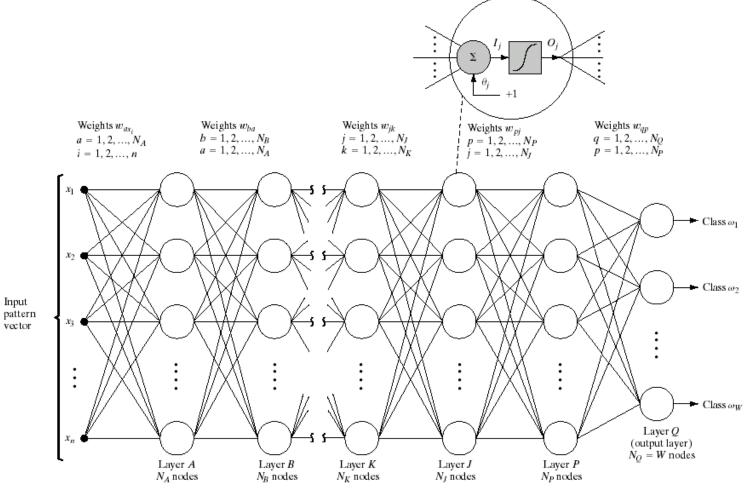
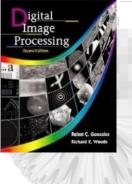


FIGURE 12.16 Multilayer feedforward neural network model. The blowup shows the basic structure of each neuron element throughout the network. The offset, θ_i , is treated as just another weight.



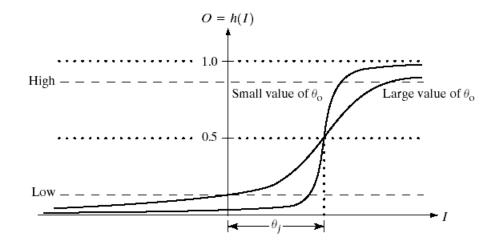


FIGURE 12.17 The sigmoidal activation function of Eq. (12.2-47).

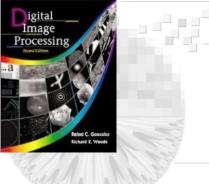




FIGURE 12.18 (a) Reference

shapes and
(b) typical noisy
shapes used in
training the
neural network of
Fig. 12.19.
(Courtesy of Dr.
Lalit Gupta, ECE
Department,
Southern Illinois
University.)







Shape 1



Shape 2



Shape 2



Shape 3



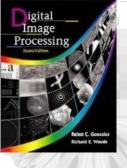
Shape 3



Shape 4



Shape 4



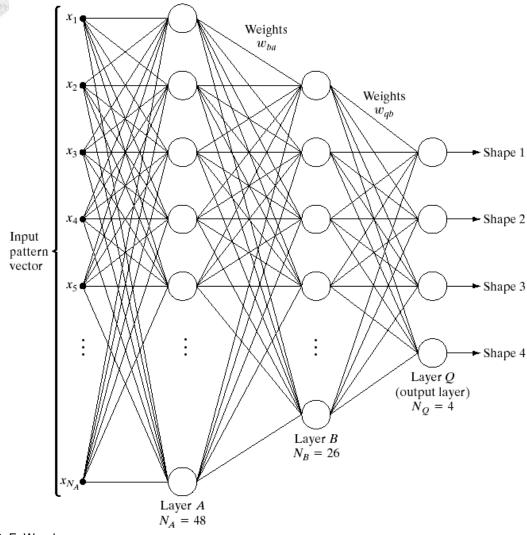
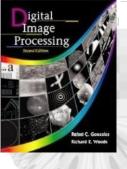


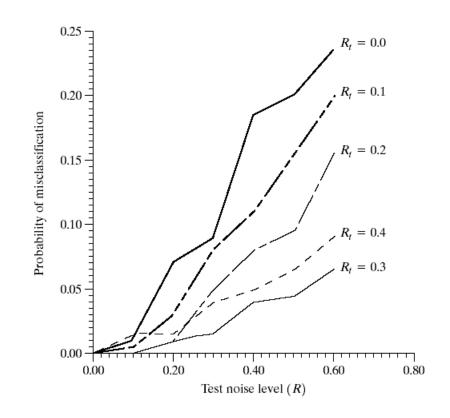
FIGURE 12.19

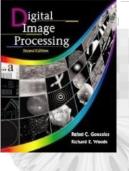
Three-layer neural network used to recognize the shapes in Fig. 12.18. (Courtesy of Dr. Lalit Gupta, ECE Department, Southern Illinois University.)



Performance of the neural network as a function of noise level. (Courtesy of Dr. Lalit Gupta, ECE Department, Southern Illinois

University.)





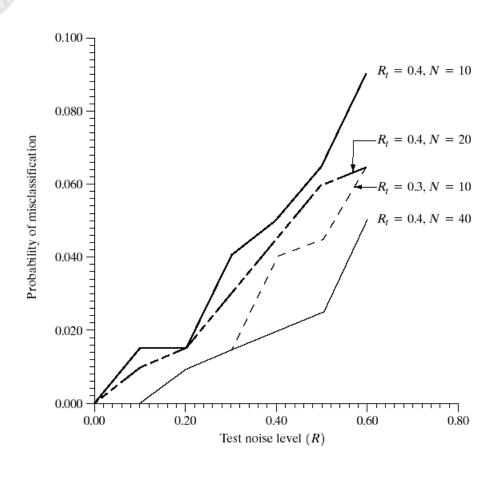
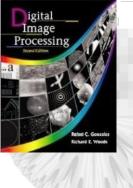
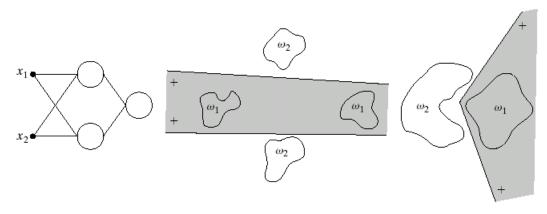


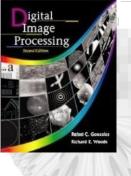
FIGURE 12.21 Improvement in performance for $R_t = 0.4$ by increasing the number of training patterns (the curve for $R_t = 0.3$ is shown for reference). (Courtesy of Dr. Lalit Gupta, ECE Department, Southern Illinois University.)





a b c

FIGURE 12.22 (a) A two-input, two-layer, feedforward neural network. (b) and (c) Examples of decision boundaries that can be implemented with this network.

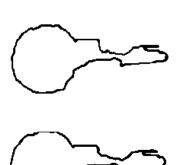


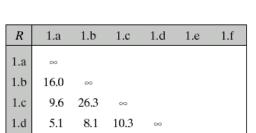
Network structure	Type of decision region	Solution to exclusive-OR problem	Classes with meshed regions	Most general decision surface shapes	
Single layer	Single hyperplane	(W ₁) (W ₂) (W ₁)	ω_2 ω_1		
Two layers	Open or closed convex regions	(W1) (W2) (W1)	ω_2 ω_1		
Three layers	Arbitrary (complexity limited by the number of nodes)	(W₁) (W2) (W1)	ω_2 ω_1		

FIGURE 12.23

Types of decision regions that can be formed by single- and multilayer feed-forward networks with one and two layers of hidden units and two inputs. (Lippman)







10.3

7.2 10.3

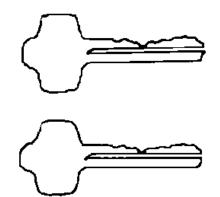
4.7

4.7

14.2

8.4

23.7

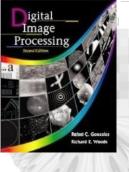


R	2.a	2.b	2.c	2.d	2.e	2.f
2.a	00					
2.b	33.5	00				
2.c	4.8	5.8	00			
2.d	3.6	4.2	19.3	00		
2.e	2.8	3.3	9.2	18.3	00	
2.f	2.6	3.0	7.7	13.5	27.0	00

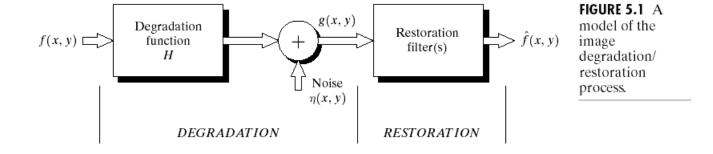


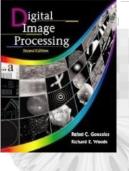
FIGURE 12.25 (a) and (b) Sample boundaries of two different object classes; (c) and (d) their corresponding polygonal approximations; (e)–(g) tabulations of *R*. (Sze and Yang.)

n		4.1				4.0
K	1.a	1.b	1.c	1.d	1.e	1.1
2.a	1.24	1.50	1.32	1.47	1.55	1.48
2.b	1.18	1.43	1.32	1.47	1.55	1.48
2.c	1.02	1.18	1.19	1.32	1.39	1.48
2.d	1.02	1.18	1.19	1.32	1.29	1.40
2.e	0.93	1.07	1.08	1.19	1.24	1.25
2.f	0.89	1.02	1.02	1.24	1.22	1.18



Digital Image Processing, 2nd ed. Image Restoration





Digital Image Processing, 2nd ed. Image Restoration

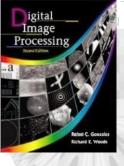
a b c d

FIGURE 5.25

Illustration of the atmospheric turbulence model. (a) Negligible turbulence. (b) Severe turbulence, k = 0.0025.(c) Mild turbulence, k = 0.001.(d) Low turbulence. k = 0.00025. (Original image courtesy of NASA.)



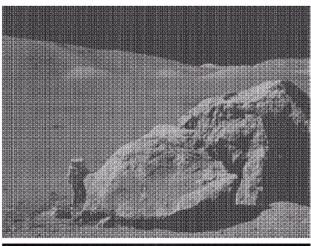


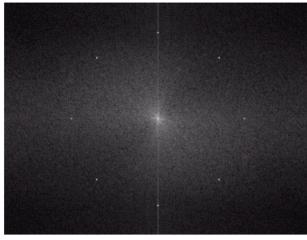


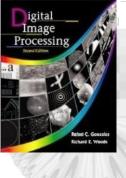
a

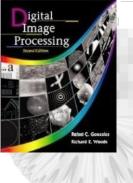
FIGURE 5.5

(a) Image corrupted by sinusoidal noise. (b) Spectrum (each pair of conjugate impulses corresponds to one sine wave). (Original image courtesy of NASA.)

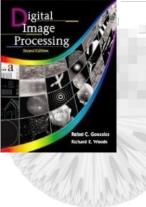








Extension to 3D and Color



Thank you!