Kernel Optimization in Pattern Recognition Using a Genetic Algorithm

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Figure 1: (a)Estimation of posterior probability by using kernel functions in the K-B method.(b)Kernel regions in the conventional K-B method and the extended K-B method.

the patterns are extracted automatically and embedded in the kernel regions. This feature extraction capability enables the extended kernel-based method to recognize patterns accurately.

Table 1: Recognition accuracies in the sonar spectrum recognition problem.

	Extended K-B	Nearest neighbor	Neural network
sample patterns	100.0%	100.0%	99.7%
test patterns	83.9%	82.7%	83.0%

References

[1] K. Fukunaga, "Introduction to Statistical Pattern Recognition," Academic Press, 1989.

SUMMARY

An extension of the conventional kernel-based method (Parzen density estimation method), which is one of fundamental pattern recognition methods [1], is proposed. In the conventional method, a single kernel function $K(\cdot)$ is used commonly in the superposition to make discrimination functions, or posterior probability functions $P(C_i)$. Figure 1(a) shows an example of the superposition in 1-dimensional space (S_i and d represent a sample pattern and a kernel region size, respectively).

In the extended method, each edge length of every kernel region is optimized individually. A difference in region between the conventional and the extended kernelbased methods in 2-dimensional space is shown in Fig. 1(b) $(d_{jk}$ in Fig. 1(b)-2 represents the *k*th edge length of the *j*th kernel region).

If there are L candidates for each edge length, the number of candidates for the *n*-dimensional kernel region is L^n . In real applications *n* is more than 50 and *L* should be about 10 or more. Thus, we have to solve the combinatorial optimization problem for 10^{50} or more candidates and this cannot be done by using a brute-force search. We therefore apply a genetic algorithm to the combinatorial optimization problem.

In our GA operation, a chromosome represents the kernel region in an *n*-dimensional pattern space, and each locus corresponds to one of the candidates (genes) for an edge length of the kernel region. We have applied the extended method to a sonar spectrum recognition problem (http://www.boltz.cs.cmu.edu) and obtained a recognition accuracy of 83.9%, which is much higher than the 62.0% obtained using the conventional kernelbased method. We have also compared the recognition accuracy of the extended K-B method with those of the nearest neighbor method and a neural network (backpropagation algorithm). The results are listed in Table 1.

We have analyzed the individually optimized kernel regions and shown that in the GA process the features in