

Argon injection number	PDF change measure for class 1	PDF change measure for class 2	Updating
2 nd	1	0.28	Class 1
3 rd	1	0.8	Classes 1 and 2
4 th	0.93	0.96	Classes 1 and 2

Table 1: PDF change measure in response to the successive argon injections of Fig 2.

Argon injection number	Misclassification error	
	Incremental classifier	Dynamic classifier
2 nd	19	19
3 rd	33	8
4 th	58	4

Table 2: Misclassification rate calculated for incremental classifier (without forgetting and updating) and dynamic classifier according to the successive argon injections of Fig 2.

4. Conclusion

In this paper, an approach to learn a classifier and to update its parameters and structure according to the changes in its environment conditions is proposed. This approach detects the changes in the classifier environment by observing the changes in its conditional probability density function during a sliding time window. When a significant change is detected, this approach updates the classifier using an incremental learning rule.

We are looking to develop an ensemble classifier approach for the learning in dynamic environments. This approach uses different classification methods to build an ensemble of classifiers. Each classifier is an expert for the classification of patterns in a particular region of the feature space. In addition, each classifier will be adapted for the monitoring of a particular type of changes (gradual, abrupt and recurring).

Acknowledgment

This work is supported by the Scientific Interest Group surveillance, safety and security of the big systems (GIS3SGS).

References

[1] B. D. Ripley, *Pattern Recognition and Neural Networks*, Cambridge University Press, Cambridge, 1996.

[2] C.W. Therrien, *Decision Estimation and Classification: An Introduction to Pattern Recognition and Related Topics*, John Wiley & Sons, New York, 1989.

[3] E. Parzen, On the estimation of a probability density function and mode, *Annals of Mathematical Statistics*, 33, 1065-1076, 1962.

[4] G. Nakhaeizadeh, C. Taylor and G. Kunisch, Dynamic Supervised Learning. Some Basic Issues and Application Aspects. *Classification and Knowledge Organization (R. Klar et O. Opitz (Eds.))*, Springer Verlag, Berlin, Heidelberg, 123-135, 1997.

[5] G.S. Srinivasan, Om Pal Singh and Prabhakar R, Leak noise detection and characterization using statistical features, *Annals of Nuclear Energy*, Vol. 27, n° 4, 2000, pp 329-343.

[6] G. Vachtsevanos, F. L. Lewis, M. Roemer, A. Hess and B. Wu, *Intelligent Fault Diagnosis and Prognosis for Engineering Systems*, John Wiley & Sons, New York, 2006.

[7] H. Frigui and R. Krishnapuram, Clustering by competitive agglomeration, *Pattern Recognition*, 30 (7), 1109-1119, 1997.

[8] H. Akaike, A new look at the statistical model identification, *IEEE Trans Autom Cont.*, 19, 716-723, 1974.

[9] L. Angstenberger, *Dynamic Fuzzy Pattern Recognition. Dissertation, Fakultät für Wirtschaftswissenschaften der Rheinisch-Westfälischen Technischen Hochschule*, Aachen, Germany, 2000.

[10] L.I. Smith, *A Tutorial on Principal Components Analysis*. Cornell University, USA, 2002.

[11] L. Cohen, M. Last and G. Avrahami, Incremental Info-Fuzzy Algorithm for Real Time Data Mining of Non-Stationary Data Streams, *TDM Workshop*, Brighton UK, 2004.

[12] M. Last, Online classification on non stationary data streams, *Intelligent Data Analysis*, 6 (2), 129-147, 2002.

[13] M. Sayed Mouchaweh, A. Devillez, G. Villerman Lecolier and P. Billaudel, Incremental learning in fuzzy pattern matching, *Fuzzy Sets and Systems*, 132 (1), 49-62, 2002.

[14] P.P Angelov, A fuzzy controller with evolving structure, *Information Sciences*, 161 (1-2), 21-35, 2004.

[15] R.O. Duda, P. E. Hart and D. G. Stork, *Pattern Classification 2nd edition*. Wiley-Interscience, 2001.

[16] R.R. Yager, An extension of the naive Bayesian classifier, *Information Sciences*, 176 (5), 577-588, 2006.

[17] S.H. Cha, Comprehensive survey on distance/similarity measures between probability density functions, *International Journal of Mathematical Models and Methods in Applied Sciences*, 4, 300-307, 2007.

[18] S. Medasani, K. Jaeseok and R. Krishnapuram, An overview of membership function generation techniques for pattern recognition, *International Journal of Approximate Reasoning*, 19, 391-417, 1998.

[19] T.M. Cover and P.E. Hart, Nearest neighbor pattern classification, *IEEE Transaction Information Theory*, 13, 21-27, 1967.

[20] V. Vapnik, *Statistical Learning Theory*, John Wiley & Sons Inc., New York, 1998.