

Reacting to Different Types of Concept Drift: The Accur

IEEE Transactions on Neural Networks and Learning Systems
25, 81-94

DOI: [10.1109/tnnls.2013.2251352](https://doi.org/10.1109/tnnls.2013.2251352)

Citation Report

#	ARTICLE	IF	CITATIONS
1	Stream Mining Using Statistical Relational Learning. , 2014, , .		2
2	The Parzen kernel approach to learning in non-stationary environment. , 2014, , .		11
3	Sleep onset detection based on Time-Varying Autoregressive models with particle filter estimation. , 2014, , .		1
4	A novel application of Hoeffding's inequality to decision trees construction for data streams. , 2014, , .		12
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