

nonlinear co-ordinate systems and non-planar image manifolds must be found. If this can be done, it will greatly increase both the computational and the empirical predictive power of abstract unsupervised learning techniques.

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