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LIST OF PUBLICATIONS

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2) K. Teeyapan, N. Theera-Umpon, and S. Auephanwiriyakul, “Ellipsoidal support vector data description,” *Neural Comput. Appl.*, vol. 28 (Suppl 1), pp. 337–347, Dec. 2017.

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