

Bibliography for Graph Classification

Kernel and Neural Computation Methods

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Abstract

Bibliography for the area of (supervised) graph classification, both graph kernel and deep learning approaches.
We only included published articles.

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