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Feature network models for proximity data : statistical inference, model selection, network representations and links with related models

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