Intelligent Assistance for Expert-Driven Subpopulation Discovery in High-Dimensional Time-Stamped Medical Data

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22.06.2020

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Based on: [[1](#ref-Niemann:ESWA2014)]

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Based on: [[2](#ref-Niemann:CBMS2017)]

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