

# **Graphs in Machine Learning**Distributed Graph Processing

Challenges and Requirements

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Partially based on material by: Rob Fergus, Nikhil Srivastav Yiannis Koutis, Joshua Batson, Daniel Spielman

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updates should propagate properly

Different choices have different impacts: for example splitting the graph according to nodes or according to edges.

Many computation models (academic and commercial) each with its pros and cons

MapReduce

MPI

Pregel

Graphlab

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https://misovalko.github.io/mva-ml-graphs.html