

# Partially Optimal Cubic Subspace Clustering

Research Project Machine Learning

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- 1 Introduction
- 1 Partial Optimality for Cubic Clique Partition Problem**
- 3 Cubic Subspace Instance Construction
- 4 Experiments and Evaluation
- 5 Research Results

# Partial Optimality for Cubic Clique Partition Problem

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Apply partial optimality conditions → solve subproblems

# Partial Optimality Algorithm

## **Partial Optimality Algorithm:**

**Input:** clustering  $y$  without fixed labels

**while** condition applied **do**

    apply subproblem-CUT-condition exhaustively

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## Reduction to subproblems:

- 1 Subproblem-CUT-condition: fix CUT labels for element pairs from different sample subsets; solve each subset as an independent problem and accumulate the results in  $c_\emptyset$ ;
- 2 JOIN-Conditions: fix JOIN labels for elements of the sample subset; add the join-cost to  $c_\emptyset$ ; solve the problem where the subset is considered as one sample;

## Subproblem-CUT and JOIN-Subset

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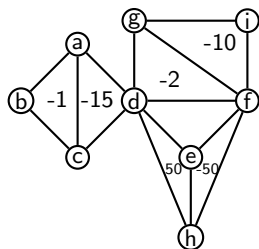
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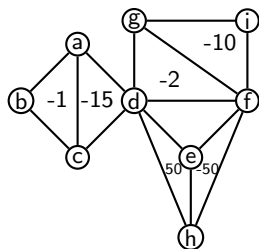


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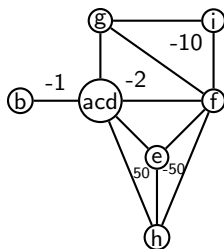
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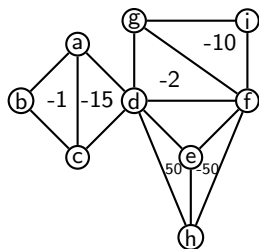


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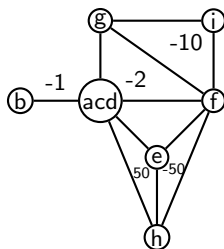
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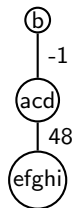
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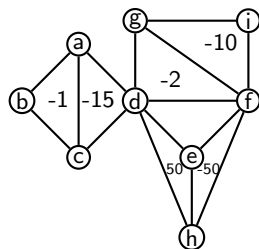


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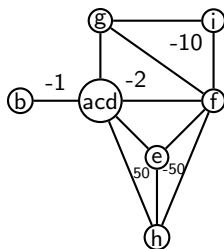
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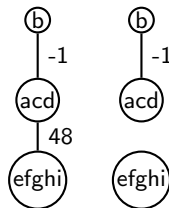
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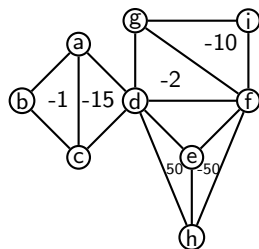


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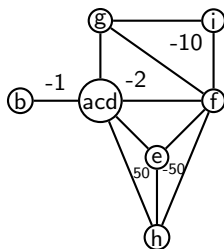
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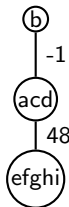
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$$c_{\emptyset} = -76$$

# JOIN-conditions

Overview of the other join-conditions (with pictures)

# CUT-conditions

Overview of the cut-conditions (with pictures)

## TODO

Class Diagram Algorithm implementation in ClusteringProblem

Features: ClusteringProblem is generally defined for all types of Cubic Clique Partition Problem (not necessarily points), cost function + sparse costs!, label computation, cut triples, logs joins and cuts! (add screenshots)