

# GRAPH PARTITIONING WITH DEMANDS: GENERALIZED GRAPH CONDUCTANCE WITH APPLICATIONS TO HIERARCHICAL CLUSTERING

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In this work, we study various graph partitioning tasks under a general demand model. In each such task, we are given a graph  $G = (V, E, c, w)$  with a capacity function  $c: E \rightarrow \mathbb{N}$  and a demand function  $w: V \times V \rightarrow \mathbb{N}$ . Our main focus is the problem of finding a cut  $(S, \bar{S})$  minimizing the quantity

$$\psi_w(S) = \frac{c(S, \bar{S})}{w(S, V) \cdot w(\bar{S}, V)}.$$

Here,  $c(S, \bar{S})$  is the cost of edges between  $S$  and the complement of  $S$ ,  $\bar{S}$ , and  $w(S, V) = w(S) + w(S, \bar{S})$  is the sum of the internal demand within  $S$ , denoted  $w(S)$ , and the demand between vertices of  $S$  and  $\bar{S}$ , denoted  $w(S, \bar{S})$ . We call  $\psi_w(S)$  the *generalized conductance* of the cut  $(S, \bar{S})$ , and the task of minimizing  $\psi_w(S)$  the GENERALIZED MINIMUM CONDUCTANCE PROBLEM. Our main contribution is an algorithm with an  $O(\log n)$ -approximation guarantee for this objective. This result is achieved via a two-way reduction: first to the well-known generalized  $k$ -multicut problem, and then to a constrained variant of the classic sparsest-cut problem, with an additional upper-bound constraint on the amount of demand that may be cut.

Moreover, we show that this result can be leveraged to obtain an  $O(\log n)$  bicriteria approximation for GRAPH PARTITIONING WITH DEMANDS, where the goal is to find a minimum-cost subset of edges  $C$  such that for every component  $H$  of  $G \setminus C$ ,  $w(H) \leq w(V)/2$ . This, in turn, yields an  $O(\log n)$ -approximation for HIERARCHICAL CLUSTERING WITH DEMANDS, the problem of finding a hierarchy of cuts that partitions the graph into increasingly refined clusters.

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