


# Improved Approximation Algorithms for Capacitated Fault-Tolerant $k$ -Center

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**Abstract** In the  $k$ -center problem, given a metric space  $V$  and a positive integer  $k$ , one wants to select  $k$  elements (centers) of  $V$  and an assignment from  $V$  to centers, minimizing the maximum distance between an element of  $V$  and its assigned center. One of the most general variants is the *capacitated  $\alpha$ -fault-tolerant  $k$ -center*, where centers have a limit on the number of assigned elements, and, if any  $\alpha$  centers fail, there is a reassignment from  $V$  to non-faulty centers. In this paper, we present a new approach to tackle fault tolerance, by selecting and pre-opening a set of backup centers, then solving the obtained residual instance. For the  $\{0, L\}$ -capacitated case, we give approximations with factor 6 for the basic problem, and 7 for the so called *conservative* variant, when only clients whose centers failed may be reassigned. Our algorithms improve on the best previously known factors of 9 and 17, respectively. Moreover, we consider the case with general capacities. Assuming  $\alpha$  is constant, our method leads to the first approximations for this case. We also derive approximations for the capacitated fault-tolerant  $k$ -supplier problem.

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## 1 Introduction

The  $k$ -center is the minimax problem in which, given a metric space  $V$  and a positive integer  $k$ , we want to choose a set of  $k$  centers such that the maximum distance from an element of  $V$  to its closest center is minimized. More precisely, the goal is to select  $S \subseteq V$  with  $|S| = k$  that minimizes

$$\max_{u \in V} \min_{v \in S} d(u, v),$$

where  $d(u, v)$  is the distance between  $u$  and  $v$ . The elements of set  $S$  are usually referred to as *centers*, and the elements of  $V$  as *clients*. The decision version of  $k$ -center appears as problem MS9 in Garey and Johnson's list of NP-complete problems [1]. It is well known that  $k$ -center has a 2-approximation, which is best possible unless  $P = NP$  [2–6].

In a typical application of  $k$ -center, set  $V$  represents the nodes of a network, and one may want to install  $k$  routers so that the network latency is minimized. Other applications have additional constraints, so variants of  $k$ -center have been considered as well. For example, the number of nodes that a router may serve might be limited. In the *capacitated  $k$ -center*, in addition to the set of selected centers, we also want to obtain an assignment from the set of clients to centers such that at most  $L_u$  clients are assigned to each center  $u$ . The number  $L_u$  is called the *capacity* of  $u$ . The first approximation for this version of the problem is due to Bar-Ilan et al. [7], who gave a 10-approximation for the particular case of uniform capacities, where there is a number  $L$  such that  $L_u = L$  for every  $u$  in  $V$ . This was improved by Khuller and Sussmann [8], who obtained a 6-approximation, and also considered the *soft* capacitated case, in which multiple centers may be opened at the same location, obtaining a 5-approximation, both results for uniform capacities.

Despite the progress in the approximation algorithms for related problems, such as the metric facility location problem, the first constant approximation for the (non-uniformly) capacitated  $k$ -center was obtained only in 2012, by Cygan et al. [9]. Differently from algorithms for the uniform case, the algorithm of Cygan et al. is based on the relaxation of a linear programming (LP) formulation. Since the natural formulation for  $k$ -center has unbounded integrality gap, a preprocessing is used, which allows considering only instances whose LP has bounded gap. The rounding uses the notion of transferring fractional values of the LP variables. They also presented an 11-approximation for the soft capacitated case. Later, An et al. [10] presented a cleaner rounding algorithm and obtained an improved approximation with factor 9 (while the previous approximation had a large constant factor, not explicitly calculated). Cygan et al. [11] also presented an algorithm for a variant of the problem with outliers. As for negative results, it has been shown that the capacitated  $k$ -center has no approximation with factor better than 3 unless  $P = NP$  [9].

Another natural variant of  $k$ -center comprises the possibility that centers may fail during operation. This was first discussed by Krumke [12], who considered the version in which clients must be connected to a given minimum number of centers. In the *fault-tolerant  $k$ -center*, for a number  $\alpha$ , we consider the possibility that any subset of centers of size at most  $\alpha$  may fail, so that a client might have to be connected to the  $(\alpha + 1)$ -th closest center. The objective is thus to minimize the maximum distance from a client to its  $\alpha + 1$  nearest centers. For the variant in which selected centers do not need to be served, Krumke [12] gave a 4-approximation, later improved to a (best possible) 2-approximation by Chaudhuri et al. [13] and Khuller et al. [14]. For the standard version, in which a client must be served even if a center is installed at the client's location, there is a 3-approximation by Khuller et al. [14], who also gave a 2-approximation for the particular case of  $\alpha \leq 2$ .

Chechik and Peleg [15] considered a common generalization of the capacitated  $k$ -center and the fault-tolerant  $k$ -center, where centers have limited capacity and may fail during operation. They defined only the uniformly capacitated version, presenting a 9-approximation. Also, they considered the case in which, after failures, only clients that were assigned to faulty centers may be reassigned. For this variant, called the conservative fault-tolerant  $k$ -center, a 17-approximation was obtained for the uniformly capacitated case. For the special case in which  $\alpha < L$ , the so called *large capacities* case, they obtained a 13-approximation.

## 1.1 Our Contributions and Techniques

We consider the *capacitated  $\alpha$ -fault-tolerant  $k$ -center* problem. Formally, an instance of this problem consists of a metric space  $V$  with corresponding distance function  $d : V \times V \rightarrow \mathbb{R}_{\geq 0}$ , non-negative integers  $k$  and  $\alpha$ , with  $\alpha < k$ , and a non-negative integer  $L_v$  for each  $v$  in  $V$ . A solution is a subset  $S$  of  $V$  with  $|S| = k$ , such that, for each  $F \subseteq S$  with  $|F| \leq \alpha$ , there exists an assignment  $\phi_F : V \rightarrow S \setminus F$  with  $|\phi_F^{-1}(v)| \leq L_v$  for each  $v$  in  $S \setminus F$ . For a given  $F$ , we denote by  $\phi_F^*$  an assignment  $\phi_F$  with minimum  $\max_{u \in V} d(u, \phi_F(u))$ . The problem's objective is to find a solution that minimizes

$$\max_{u \in V, F \subseteq V: |F| \leq \alpha} d(u, \phi_F^*(u)).$$

We also consider the *capacitated conservative  $\alpha$ -fault-tolerant  $k$ -center*. In this variant, in addition to the set  $S$ , a solution comprises an initial assignment  $\phi_0$ . We require that the assignment  $\phi_F$  for a failure scenario  $F$  differs from  $\phi_0$  only for vertices assigned by  $\phi_0$  to centers in  $F$ . Precisely, given  $F \subseteq S$  with  $|F| \leq \alpha$ , we say that an assignment  $\phi_F$  is *conservative* (with respect to  $\phi_0$ ) if  $\phi_F(u) = \phi_0(u)$  for every  $u \in V$  with  $\phi_0(u) \notin F$ . A solution for the problem is a pair  $(S, \phi_0)$  such that, for each  $F \subseteq S$  with  $|F| \leq \alpha$ , there exists a conservative assignment  $\phi_F$ . The objective function is defined analogously.

Our major technical contribution is a new strategy to deal with the fault-tolerant and capacitated problems. Namely, we solve the considered problems in two phases. In the first phase, we identify clusters of vertices where an optimal solution must

install a minimum of  $\alpha$  centers. For each cluster, we carefully select  $\alpha$  of its vertices, and pre-open them as centers. These  $\alpha$  centers will have enough *backup* capacity so that, in the case of failure events, the unused capacity of all pre-opened centers will be sufficient to obtain a reassignment for all clients. While the  $\alpha$  guessed centers of a cluster may not correspond to centers in an optimal solution, we carefully select elements that are near to centers of an optimal solution, so that our choice leads to an approximate solution. In the second phase, we are left with a residual instance, where part of a solution is already known. For the conservative case, obtaining the remaining centers of a solution may be reduced to the non-fault-tolerant variant. For the non-conservative case, we can make stronger assumptions over the input and the solution so that the task of obtaining a fault-tolerant solution is simplified.

A good feature of the presented approach is that it can be used in combination with different methods and algorithms, and can be applied to different versions of the problem. Indeed, we obtain approximations for both the conservative and non-conservative variants of the capacitated fault-tolerant  $k$ -center. Moreover, each of the obtained approximations uses novel and specific techniques that are of particular interest. For the conservative variant, we present elegant combinatorial algorithms that reduce the problem to the non-fault-tolerant case. For the non-conservative variant, our algorithms are based on the rounding of a new LP formulation for the problem, and apply some rounding techniques by An et al. [10]. Interestingly, we use the set of pre-opened centers to obtain a partial solution for the LP variables with integral values. We hope that other problems can benefit from similar techniques.

## 1.2 Obtained Approximations and Paper Organization

The conservative variant is considered in Sects. 3 and 4. In Sect. 3, we present a 7-approximation for the  $\{0, L\}$ -capacitated conservative  $\alpha$ -fault-tolerant  $k$ -center. This is the special case of the problem where the capacities are either 0 or  $L$ , for some  $L$ . Notice that this generalizes the uniformly capacitated case, when all capacities are equal to  $L$ . This result improves on the previously known factors of 17 and 13 by Chechik and Peleg [15], that apply to particular cases with uniform capacities, and uniform large capacities, respectively. In Sect. 4, we study the case of general capacities, and present a  $(9 + 6\alpha)$ -approximation when  $\alpha$  is constant. To the best of our knowledge, this is the first approximation for the problem with arbitrary capacities.

For the non-conservative variant, our algorithms are based on the rounding of a new LP formulation, and are described in Sects. 5 and 6. First we consider the case of arbitrary capacities in Sect. 5. We present the LP formulation, and give a 10-approximation when  $\alpha$  is constant. Once again, this is the first approximation for the problem with arbitrary capacities. In Sect. 6, the rounding algorithm is adapted for the  $\{0, L\}$ -capacitated fault-tolerant  $k$ -center, for which we obtain a 6-approximation with  $\alpha$  being part of the input. This factor matches the best known factor for the problem without fault tolerance [8, 10], and improves on the best previously known algorithm for the fault-tolerant version, which achieves factor 9 for the uniformly capacitated case [15].

**Table 1** Summary of the obtained approximation factors for the  $k$ -center problem

Version	Capacities	Value of $\alpha$	Previous	This paper
Conservative	Uniform	Given in the input	17 [15]	7
Conservative	Arbitrary	Fixed	–	$9 + 6\alpha$
Non-conservative	Uniform	Given in the input	9 [15]	6
Non-conservative	Arbitrary	Fixed	–	10

In Sect. 7, we apply our technique to the  $k$ -supplier problem. This is a generalization of  $k$ -center where one is given a set of clients, a set of candidate locations, and an integer  $k$ , and the goal is to select  $k$  of the locations to install facilities and serve each of the clients. The objective is to minimize the maximum distance between a client and its assigned facility.

Our strategy for the problems with arbitrary capacities is to reduce them to related feasibility problems. In Sect. 8, we show that these related feasibility problems are coNP-hard. The algorithms we have for them run in time  $|V|^{\Theta(\alpha)}$ , so the approximations for the corresponding variants of  $k$ -center or  $k$ -supplier are polynomial only when  $\alpha$  is fixed.

A summary of the results for  $k$ -center is given in Table 1. A similar table for  $k$ -supplier is given in Sect. 7.

## 2 Preliminaries

In what follows, we will write  $\kappa$ -center instead of  $k$ -center since some of our algorithms create instances with different values of  $k$ , yet all of them refer the same problem.

Let  $G = (V, E)$  be an undirected and unweighted graph. We denote by  $d_G$  the metric induced by  $G$ , that is, for  $u$  and  $v$  in  $V$ , let  $d_G(u, v)$  be the length of a shortest path between  $u$  and  $v$  in  $G$ . For given nonempty sets  $A, B \subseteq V$ , we define  $d_G(A, B) = \min_{a \in A, b \in B} d_G(a, b)$ . Also, for  $a \in V$ , we may write  $d_G(a, B)$  instead of  $d_G(\{a\}, B)$ .

For an integer  $\ell$ , we let  $N_G^\ell(u) = \{v \in V : d_G(u, v) \leq \ell\}$ . For a subset  $U \subseteq V$ , let  $N_G^\ell(U) = \bigcup_{u \in U} N_G^\ell(u)$ . We may omit the superscript  $\ell$  when  $\ell = 1$ , and the subscript  $G$  when the graph is clear from the context. Thus  $N(v)$  is the set of neighbors of  $v$  plus  $v$  itself. Also, we define the (power) graph  $G^\ell = (V, E^\ell)$ , where  $\{u, v\} \in E^\ell$  if  $v \in N^\ell(u) \setminus \{u\}$ . For a directed graph  $G$ , we define  $d_G(u, v)$  as the length of a shortest directed path from  $u$  to  $v$  in  $G$ , and define  $N_G^\ell(u)$  similarly.

### 2.1 Reduction to the Unweighted Case

As it is standard for the  $\kappa$ -center problem, we will use the bottleneck method [5], so that we can consider the case in which the metric space is induced by an unweighted undirected graph. Suppose we have an algorithm that, given an unweighted graph, either produces a distance- $r$  solution for the unweighted problem (that is, one in which each vertex is assigned to a center at a distance at most  $r$ ), or a certificate

that no distance-1 solution exists. We may then use this algorithm to obtain an  $r$ -approximation for the general metric case.

Let  $V$  be a metric space associated with distance function  $d : V \times V \rightarrow \mathbb{R}_{\geq 0}$ . For a certain number  $\tau$  in  $\mathbb{R}_{\geq 0}$ , we consider the *threshold graph* defined as  $G_{\leq \tau} = (V, E_{\leq \tau})$ , where  $E_{\leq \tau} = \{\{u, v\} : d(u, v) \leq \tau\}$ . Next we consider the values of  $d(u, v)$  for  $(u, v)$  in  $V^2$ , in increasing order. For each  $\tau$  in this ordering, we obtain  $G_{\leq \tau}$ , and use the algorithm for the unweighted case; we stop when the algorithm fails to provide a negative certificate, and return the obtained solution. Notice that there must be a distance-1 solution for  $G_{\leq \text{OPT}}$ , where OPT denotes the optimum value for the problem. Since OPT is in the considered ordering for  $\tau$ , the algorithm always stops, and returns a solution for some  $\tau \leq \text{OPT}$ , so we obtain a solution for the original problem of cost at most  $r \cdot \tau \leq r \cdot \text{OPT}$ . Hence, from now on, we assume that an unweighted graph  $G = (V, E)$  is given, and that the goal is to either obtain a certificate that no distance-1 solution exists, or return a distance- $r$  solution.

## 2.2 Preprocessing and Reduction to the Connected Case

We also may assume without loss of generality that  $G$  is connected [8, 9, 15]. If this is not the case, we may proceed as follows. Suppose there is an algorithm that, given a connected graph  $\tilde{G}$  and an integer  $\tilde{k}$ , produces a distance- $r$  solution with  $\tilde{k}$  vertices, or gives a certificate that no distance-1 solution with  $\tilde{k}$  vertices exists. Now, consider a given arbitrary unweighted graph  $G$ , and a given integer  $k$ . We decompose  $G$  into its connected components, say  $G_1, \dots, G_t$ . For each connected component  $G_i$ , with  $1 \leq i \leq t$ , we run the algorithm for each  $\tilde{k} = \alpha + 1, \dots, k$  and find the minimum value  $k_i$ , if any, for which the algorithm obtains a distance- $r$  solution. As the failure set is arbitrary, in the worst case all faulty centers might be in the same component. If, for some  $G_i$ , there is no distance-1 solution with  $k$  centers or if  $k_1 + \dots + k_t > k$ , then clearly there is no distance-1 solution for  $G$  with  $k$  centers; otherwise, conjoining the solutions obtained for each component leads to a distance- $r$  solution for  $G$  with no more than  $k$  centers, and this solution is tolerant to the failure of  $\alpha$  centers. From now on, we will assume that  $G$  is connected.

## 3 $\{0, L\}$ -Capacitated Conservative Fault-Tolerant $k$ -Center

After the occurrence of a failure, a distance-1 conservative solution has to reassign each unserved client to an open center in its vicinity with available capacity. This requires some kind of “local available center capacity”, to be used as backup. The next definition describes a set of vertices that are good candidates to be opened as backup centers. This set can be partitioned into clusters of at most  $\alpha$  vertices, with the clusters sufficiently apart from each other. The idea is that failures in the vicinity of one of these clusters do not affect centers in the other clusters. More precisely, the vicinities of different clusters do not intersect; therefore, in a distance-1 conservative solution, any client that is assigned to a center in a certain cluster cannot be reassigned to a center in the vicinity of any of the other clusters.

**Definition 1** Consider a graph  $G = (V, E)$  and non-negative integers  $\alpha$  and  $\ell$ . A set  $W$  of vertices of  $G$  is  $(\alpha, \ell)$ -independent if it can be partitioned into sets  $C_1, \dots, C_t$ , such that  $|C_i| \leq \alpha$  for  $1 \leq i \leq t$ , and  $d(C_i, C_j) \geq \ell$  for  $1 \leq i < j \leq t$ .

In what follows, we denote by  $(G, k, L, \alpha)$  an instance of the capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center as obtained from Sect. 2. We say that  $(G, k, L, \alpha)$  is feasible if there exists a distance-1 solution for it.

**Lemma 1** Let  $(G, k, L, \alpha)$  be a feasible instance of the capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center, and let  $(S^*, \phi_0^*)$  be a corresponding distance-1 solution. If  $W \subseteq S^*$  is an  $(\alpha, 5)$ -independent set in  $G$ , then  $(G, k - |W|, L')$  is feasible for the capacitated  $\kappa$ -center, where  $L'_u = 0$  for  $u$  in  $W$ , and  $L'_u = L_u$  otherwise.

*Proof* Since  $W$  is  $(\alpha, 5)$ -independent, there must be a partition  $C_1, \dots, C_t$  of  $W$  such that  $d(C_i, C_j) \geq 5$  for any pair  $i, j$ , with  $1 \leq i < j \leq t$ . Also, each part  $C_i$  has at most  $\alpha$  vertices, and thus there exists a conservative assignment  $\phi_{C_i}^*$  with  $(\phi_{C_i}^*)^{-1}(C_i) = \emptyset$ . Therefore,  $\phi_{C_i}^*$  is a distance-1 solution for the  $(G, k - |C_i|, L^i)$  instance of the capacitated  $\kappa$ -center problem, where  $L_u^i = 0$  for  $u$  in  $C_i$ , and  $L_u^i = L_u$  otherwise. Moreover, as  $\phi_0^*$  is conservative,  $\phi_{C_i}^*$  differs from  $\phi_0^*$  only in  $(\phi_0^*)^{-1}(C_i)$ .

So, if a center  $u$  in  $S^*$  is such that  $(\phi_0^*)^{-1}(u) \neq (\phi_{C_i}^*)^{-1}(u)$ , then  $u \in N^2(C_i)$ . As  $W$  is  $(\alpha, 5)$ -independent,  $N^2(C_i) \cap N^2(C_j) = \emptyset$  for every  $j \in [t] \setminus \{i\}$ , where  $[t] = \{1, 2, \dots, t\}$ . Let  $\psi$  be an assignment such that, for each client  $v$ ,

$$\psi(v) = \begin{cases} \phi_{C_i}^*(v) & \phi_0^*(v) \in C_i \text{ for some } i \text{ in } [t], \\ \phi_0^*(v) & \text{otherwise.} \end{cases}$$

Therefore, set  $\psi^{-1}(u)$  is empty if  $u \in W$ ; is  $(\phi_{C_i}^*)^{-1}(u)$  if there exists  $i \in [t]$  such that  $u \in N^2(C_i) \setminus C_i$ ; and is  $(\phi_0^*)^{-1}(u)$  otherwise. This means that, for  $L'$  as in the statement of the lemma,  $|\psi^{-1}(u)| \leq L'_u$  for every  $u$ , and so  $(S^*, \psi)$  is a solution for the  $(G, k - |W|, L')$  instance of the capacitated  $\kappa$ -center problem.  $\square$

A set of vertices  $A \subseteq V$  is 7-independent in  $G$  if every pair of vertices in  $A$  is at distance at least 7 in  $G$ . This definition was also used by Chechik and Peleg [15] and, as we will show, such a set is useful to obtain an  $(\alpha, 5)$ -independent set in  $G$ .

**Lemma 2** Let  $A$  be a 7-independent set in  $G$ , for each  $a$  in  $A$ , let  $B(a)$  be any set of  $\alpha$  vertices in  $N(a)$ , and let  $B = \bigcup_{a \in A} B(a)$ . If  $(G, k, L, \alpha)$  is feasible for the capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center, then  $(G, k - |B|, L')$  is feasible for the capacitated  $\kappa$ -center, where  $L'_u = 0$  for  $u$  in  $B$ , and  $L'_u = L_u$  otherwise.

*Proof* Let  $(S^*, \phi_0^*)$  be a solution for  $(G, k, L, \alpha)$ . For each  $a \in A$ , there must be at least  $\alpha$  centers in  $S^* \cap N(a)$ . Let  $W(a)$  be the union of  $S^* \cap B(a)$  and other  $\alpha - |S^* \cap B(a)|$  centers in  $S^* \cap N(a)$ . Let  $W = \bigcup_{a \in A} W(a)$ . Since  $A$  is 7-independent,  $N^3(a)$  and  $N^3(b)$  are disjoint for any two  $a$  and  $b$  in  $A$ , and so  $N^2(W(a)) \cap N^2(W(b)) = \emptyset$ . Thus,  $W$  is  $(\alpha, 5)$ -independent.

Now let  $L''$  be such that  $L''_u = 0$  if  $u \notin S^*$ , and  $L''_u = L_u$  otherwise. Observe that the instance  $(G, k, L'', \alpha)$  is feasible (as we only set to zero the capacities of non-centers).

By Lemma 1, the instance  $(G, k - |W|, L''')$  is feasible, where  $L''_u = 0$  if  $u \in W$ , and  $L''_u = L''_u$  otherwise. Notice that  $L'_u \geq L''_u$  for every  $u$ , and  $|B| = |W|$ . Therefore, since  $(G, k - |W|, L''')$  is feasible, so is  $(G, k - |B|, L')$ .  $\square$

Now we present a 7-approximation for the  $\{0, L\}$ -capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center. For this case, rather than using the capacity function, it is convenient to consider the subset of vertices with capacity  $L$ , that is denoted by  $V^L$ . We denote by  $(G, k, V^L, \alpha)$  and by  $(G, k, V^L)$  instances of the fault-tolerant and non-fault-tolerant versions. The steps are detailed in Algorithm 1, where ALG denotes an approximation algorithm for the  $\{0, L\}$ -capacitated  $\kappa$ -center.

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**Algorithm 1:**  $\{0, L\}$ -capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center.

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**Input:** connected graph  $G, k, V^L$ , and  $\alpha$

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1   $A \leftarrow$  a maximal 7-independent vertex set in  $G$ 
2  foreach  $a \in A$  do
3    if  $|N(a) \cap V^L| < \alpha$  then return FAILURE
4     $B(a) \leftarrow \alpha$  vertices chosen arbitrarily in  $N(a) \cap V^L$ 
5  end
6   $B \leftarrow \bigcup_{a \in A} B(a)$ 
7  if ALG( $G, k - |B|, V^L \setminus B$ ) returns FAILURE then return FAILURE
8  Let  $(S, \phi)$  be the solution returned by ALG( $G, k - |B|, V^L \setminus B$ )
9  return  $(S \cup B, \phi)$ 
    
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**Theorem 1** *If ALG is a  $\beta$ -approximation for the  $\{0, L\}$ -capacitated  $\kappa$ -center, then Algorithm 1 is a  $\max\{7, \beta\}$ -approximation for the  $\{0, L\}$ -capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center.*

*Proof* Consider an instance  $(G, k, V^L, \alpha)$  of the  $\{0, L\}$ -capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center problem, with  $G = (V, E)$ . Let  $A, B(a)$  for  $a$  in  $A$ , and  $B$  be as defined in Algorithm 1 with  $(G, k, V^L, \alpha)$  as input. Assume that  $(G, k, V^L, \alpha)$  is feasible. Since  $A$  is 7-independent, by Lemma 2, the instance  $(G, k - |B|, V^L \setminus B)$ , where we set to zero the capacities of all vertices in  $B$ , is also feasible for the  $\{0, L\}$ -capacitated  $\kappa$ -center problem. This means that, if Algorithm 1 executes Line 7, then the given instance is indeed infeasible. On the other hand, if ALG returns a solution  $(S, \phi)$ , then, since  $|S| \leq k - |B|$ , the size of  $S \cup B$  is at most  $k$ , and  $\phi$  is a valid initial center assignment. Moreover,  $\phi$  is such that: (1) each vertex  $u$  is at distance at most  $\beta$  from  $\phi(u)$ ; and (2) no vertex is assigned to  $B$ .

Let  $F \subseteq S \cup B$  with  $|F| = \alpha$  be a failure scenario. We describe a conservative center reassignment for  $(S \cup B, \phi)$ . We only need to reassign vertices initially assigned to centers in  $F \setminus B$  (as no vertex was assigned to a vertex in  $B$ ). Thus, at most  $L|F \setminus B|$  vertices need to be reassigned. For each such vertex  $u$ , we can choose  $a \in A$  at distance at most 6 from  $u$  (as  $A$  is maximal), and let  $\tilde{\phi}(u) = a$ . Then, for each  $a \in A$ , and for each  $u$  with  $\tilde{\phi}(u) = a$ , reassign  $u$  to some non-full center of  $B(a) \setminus F$ . Notice that  $B(a) \setminus F$  can absorb all reassigned vertices. Indeed, the available capacity of  $B(a) \setminus F$  before the failure event is  $L|B(a) \setminus F| = L|F \setminus B(a)| \geq L|F \setminus B|$ , where



we used  $|B(a)| = |F| = \alpha$ . Since for a reassigned vertex  $u$ ,  $d(u, \tilde{\phi}(u)) \leq 6$ , and  $u$  is reassigned to some center  $v \in N(\tilde{\phi}(u))$ , the distance between  $u$  and  $v$  is at most 7. Also, if a vertex  $u$  was not reassigned, then the distance to its center is at most  $\beta$ .  $\square$

Now, using the 6-approximation by An et al. [10] for the  $\{0, L\}$ -capacitated  $\kappa$ -center, we obtain the following.

**Corollary 1** *Algorithm 1 using as ALG the algorithm by An et al. [10, Theorem 11] for the  $\{0, L\}$ -capacitated  $\kappa$ -center is a 7-approximation for the  $\{0, L\}$ -capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center.*

## 4 Capacitated Conservative Fault-Tolerant $k$ -Center

In this section, we consider the capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center. Recall that this is the case in which capacities may be arbitrary. An instance of this problem is denoted by  $(G, k, L, \alpha)$  for some  $G = (V, E)$  and  $L : V \rightarrow \mathbb{Z}_{\geq 0}$ . Under the assumption that  $\alpha$  is bounded by a constant, we present the first approximation for the problem.

In the  $\{0, L\}$ -capacitated case, each vertex assigned to a faulty center could be reassigned to a non-faulty center in  $B(a)$ , for an arbitrary nearby element  $a$  of a 7-independent set  $A$ . Each  $B(a)$  could absorb *all* reassigned vertices. With arbitrary capacities, the set  $B$  of pre-opened centers must be obtained much more carefully, as the capacities of non-zero-capacitated vertices are not necessarily all the same. Once the set  $B$  of backup centers is selected, one needs to ensure that the residual instance of the capacitated  $\kappa$ -center problem is feasible. In Sect. 3, an  $(\alpha, 5)$ -independent set is obtained from  $A$ , and Lemma 1 is used. This lemma is valid for arbitrary capacities, so it is useful here as well. To obtain an  $(\alpha, 5)$ -independent set from  $B$ , we make sure that  $B$  can be partitioned in such a way that any two parts are at least at distance 7. This is done by Algorithm 2, where ALG denotes an approximation for the capacitated  $\kappa$ -center problem.

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**Algorithm 2:** capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center, fixed  $\alpha$ .

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**Input:** connected graph  $G = (V, E)$ ,  $k$ , and  $L : V \rightarrow \mathbb{Z}_{\geq 0}$

```

1 foreach  $u \in V$  do
2   if  $L_u > |V|$  then  $L_u \leftarrow |V|$ 
3 end
4  $B \leftarrow \emptyset$ 
5 while there is a set  $U \subseteq V$  with  $|U| \leq \alpha$  and  $L(U) > L(B \cap N^6(U))$  do
6    $B \leftarrow (B \setminus N^6(U)) \cup U$ 
7 end
8 foreach  $u \in V$  do
9   if  $u \in B$  then  $L'_u \leftarrow 0$  else  $L'_u \leftarrow L_u$ 
10 end
11 if  $\text{ALG}(G, k - |B|, L')$  returns FAILURE then return FAILURE
12 Let  $(S, \phi)$  be the solution returned by  $\text{ALG}(G, k - |B|, L')$ 
13 return  $(S \cup B, \phi)$ 

```

---

Algorithm 2 is polynomial in the size of  $G$ ,  $k$ , and  $L$ . The test in Line 5 can be implemented by finding a set  $U \subseteq V$  with  $|U| \leq \alpha$  that minimizes  $L(B \cap N^6(U)) - L(U)$  (note that this is a particular case of minimizing a submodular function with cardinality constraint). If, for an arbitrary  $\alpha$ , there were a polynomial-time algorithm for finding such a set  $U$ , then Algorithm 2 would be polynomial also in  $\alpha$ . In Sect. 8, we give evidence that such algorithm only exists if  $P = NP$ . When  $\alpha$  is fixed, we may enumerate the sets  $U$  in polynomial time. In the following, we show that Algorithm 2 is an approximation algorithm for the capacitated conservative fault-tolerant  $\kappa$ -center assuming that  $\alpha$  is fixed.

The next lemma is analogous to Lemma 2, but it applies to the case with general capacities.

**Lemma 3** *Let  $B$  be the set of vertices obtained by Algorithm 2 after the execution of Lines 4–7. If the instance  $(G, k, L, \alpha)$  is feasible for the capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center, then the instance  $(G, k - |B|, L')$  is feasible for the capacitated  $\kappa$ -center, where  $L'_u = 0$  for  $u$  in  $B$ , and  $L'_u = L_u$  otherwise.*

*Proof* Recall Definition 1: a set of vertices is  $(\alpha, \ell)$ -independent if it can be partitioned into sets  $C_1, \dots, C_t$  such that  $|C_i| \leq \alpha$  for  $1 \leq i \leq t$ , and  $d(C_i, C_j) \geq \ell$  for  $1 \leq i < j \leq t$ . Let us argue that  $B$  is  $(\alpha, 7)$ -independent.

Let  $t$  be the number of components of  $G^6[B]$  and take each  $C_i$  to be the vertex set of one of the components of  $G^6[B]$ . Let us argue that  $|C_i| \leq \alpha$  for every  $i$  with  $1 \leq i \leq t$ . Suppose, for a contradiction, that  $|C_i| > \alpha$  for some  $i$  and let  $U'$  be the vertices in  $C_i$  that were inserted in  $B$  in the last iteration of Line 6 when a vertex in  $C_i$  was added to  $B$ . Clearly  $|U'| \leq \alpha$ . Since  $C_i$  corresponds to a connected component in  $G^6[B]$  and  $|C_i| > \alpha$ , there must be a vertex contained in  $C_i \setminus U'$  in  $N^6(U') \cap B$  at this execution of Line 6, but then this vertex would have been removed from  $B$ , a contradiction. So  $B$  is indeed  $(\alpha, 7)$ -independent.

Now, for each  $i$  with  $1 \leq i \leq t$ , choose an arbitrary element  $a_i$  in  $C_i$ . (Note that the set  $A = \{a_1, \dots, a_t\}$  is 7-independent in  $G$ .)

Consider a solution  $(S^*, \phi_0^*)$  for  $(G, k, L, \alpha)$  and observe that, for each  $i$  with  $1 \leq i \leq t$ , there must be at least  $\alpha + 1$  centers in  $S^* \cap N(a_i)$ . So let  $W_i$  be the union of  $C_i \cap S^*$  and other  $|C_i \setminus S^*|$  centers in  $S^* \cap N(a_i)$ . The set  $W_i$  is well defined, as  $|C_i| \leq \alpha < |S^* \cap N(a_i)|$ . Moreover,  $|W_i| = |C_i|$  and  $W_i \subseteq N(C_i)$ .

Let  $W = \bigcup_{i=1}^t W_i$  and note that  $|W| = |B|$ . For each pair  $i, j$  with  $1 \leq i < j \leq t$  we have that  $d(W_i, W_j) \geq 5$ , because  $B$  is  $(\alpha, 7)$ -independent and thus  $d(C_i, C_j) \geq 7$ . Hence,  $W$  is  $(\alpha, 5)$ -independent.

Let  $L''$  be such that  $L''_u = 0$  if  $u \notin S^*$ , and  $L''_u = L_u$  otherwise. Observe that the instance  $(G, k, L'', \alpha)$  is feasible (as we only set to zero the capacities of non-centers). By Lemma 1, the instance  $(G, k - |W|, L''')$  is feasible, where  $L'''_u = 0$  if  $u \in W$ , and  $L'''_u = L''_u$  otherwise. Notice that  $L'_u \geq L'''_u$  for every  $u$ , and  $|B| = |W|$ . Therefore, since  $(G, k - |W|, L''')$  is feasible, so is  $(G, k - |B|, L')$ .  $\square$

**Theorem 2** *If ALG is a  $\beta$ -approximation for the capacitated  $\kappa$ -center, then Algorithm 2 is a  $(\beta + 6\alpha)$ -approximation for the capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center with fixed  $\alpha$ .*

*Proof* Let  $(G, k, L, \alpha)$  be an instance of the capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center. No center can have more than  $|V|$  clients assigned to it, so Line 2 does not affect a solution.

Since  $\alpha$  is fixed, each execution of Line 5 takes time polynomial in  $|V|$ . Also, each execution of Line 6 increases the value of  $L(B)$  by at least one. But  $L(B)$  is an integer, starts from 0, and is at most  $|V|^2$ , because each vertex capacity is at most  $|V|$  after executing Line 2. Thus, the number of iterations is quadratic in  $|V|$ , and each one takes time polynomial in  $|V|$ . Finally, as ALG is a polynomial-time algorithm, we conclude that Algorithm 2 is polynomial.

By Lemma 3, we know that, if ALG returns FAILURE in Line 11, then the instance  $(G, k, L, \alpha)$  is infeasible for the capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center. On the other hand, if ALG returns a solution  $(S, \phi)$ , then  $(S \cup B, \phi)$  is a valid set of centers and initial attribution for our problem, and is such that each vertex  $u$  is at distance at most  $\beta$  from  $\phi(u)$ . To complete our proof, we argue next that, for each failure scenario, each client  $u$  of a faulty center can be reassigned to a center at distance at most  $\beta + 6\alpha$  from  $u$ , and no center has its capacity exceeded by the reassignment.

Consider a failure scenario  $F \subseteq V$  with  $|F| = \alpha$ . We define next a flow network  $(H, c, s, t)$ , with source  $s$  and sink  $t$ , in which a maximum flow from  $s$  to  $t$  provides a valid distance- $(\beta + 6\alpha)$  reassignment for the clients of centers in  $F$  (see Fig. 1). Network graph  $H = (V_H, E_H)$  is such that the set  $V_H$  of vertices is comprised of

- a source  $s$  and a terminal  $t$ ,
- a copy of each  $y$  in  $\phi^{-1}(F)$ ,
- a copy of each  $v$  in  $F$ ,
- a copy of each  $u$  in  $B$ , denoted by  $\bar{u}$ .

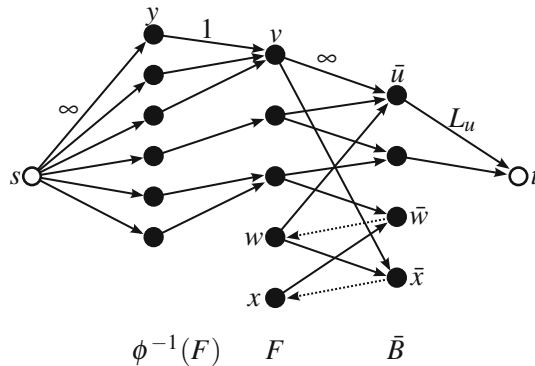
Denote by  $\bar{B} = \{\bar{u} : u \in B\}$ , and  $\bar{F} = \{\bar{w} \in \bar{B} : w \in F\}$ ; also, for  $\bar{u} \in \bar{B}$ , let  $L_{\bar{u}} = L_u$ . The set  $E_H$  of arcs is comprised of

- for each  $y$  in  $\phi^{-1}(F)$ , an arc  $(s, y)$  with capacity  $c(s, y) = \infty$ ,
- for each  $v$  in  $F$  and each  $y$  in  $\phi^{-1}(v)$ , an arc  $(y, v)$  with  $c(y, v) = 1$ ,
- for each  $v$  in  $F$  and each  $u$  in  $B \cap N_G^0(v)$ , an arc  $(v, \bar{u})$  with  $c(v, \bar{u}) = \infty$ ,
- for each  $u$  in  $B \setminus F$ , a forward arc  $(\bar{u}, t)$  with capacity  $c(\bar{u}, t) = L_u$ , and,
- for each  $w$  in  $B \cap F$ , a reverse arc  $(\bar{w}, w)$  with  $c(\bar{w}, w) = \infty$ .

Let  $C$  be a minimum capacity  $s$ – $t$  cut in  $H$  and  $(X, Y)$  be the corresponding partition of the vertices, with  $s \in X$  and  $t \in Y$ . Define  $U = X \cap F$ . Since each arc with tail in  $F$  has infinite capacity, we have that  $N_H(U) \cap \bar{B} \subseteq X \cap \bar{B}$ . Also define  $\bar{Q} = X \cap \bar{F}$ , and  $Q = \{w : \bar{w} \in \bar{Q}\}$ . As each arc of the form  $(\bar{w}, w)$  has infinite capacity, if  $\bar{w} \in \bar{Q}$ , then  $w \in U$ , and thus  $Q \subseteq U$ .

Since only arcs from  $\phi^{-1}(F)$  to  $F$  and from  $\bar{B} \setminus \bar{F}$  to  $t$  have finite capacities, they are the only ones that can be in a minimum cut set, and thus the capacity of  $C$  can be expressed as

$$\begin{aligned} c(C) &= |\phi^{-1}(F \setminus X)| + L((\bar{B} \setminus \bar{F}) \cap X) \\ &\geq |\phi^{-1}(F)| - |\phi^{-1}(U)| + L(N_H(U) \cap \bar{B}) - L(\bar{Q}) \\ &\geq |\phi^{-1}(F)| - L(U) + L(Q) + L(U) - L(\bar{Q}) = |\phi^{-1}(F)|. \end{aligned}$$



**Fig. 1** The flow network defined in terms of  $L$ ,  $B$ ,  $\phi$ , and  $F$

The first inequality comes from the definition of  $U$  and  $\bar{Q}$ , and from  $N_H(U) \cap \bar{B} \subseteq X \cap \bar{B}$ . The second inequality holds because  $\phi$  does not assign any vertex to  $Q \subseteq U$ , so  $|\phi^{-1}(U)| \leq L(U) - L(Q)$ ; and since from the loop starting at Line 5 of Algorithm 2, we have that  $L(U) \leq L(N_G^6(U) \cap B) = L(N_H(U) \cap \bar{B})$ .

Hence the value of a maximum integer flow on  $H$  is exactly  $|\phi^{-1}(F)|$ , and thus every arc from  $\phi^{-1}(F)$  to  $F$  has flow exactly 1. It is straightforward to obtain an assignment  $\psi : \phi^{-1}(F) \rightarrow B \setminus F$ . For each vertex  $y$  in  $\phi^{-1}(F)$ , let  $\psi(y) = u$ , where  $\bar{u}$  is the center in  $\bar{B} \setminus \bar{F}$  that receives the unit of flow going through  $y$  (for example, in Fig. 1, a unit of flow could traverse a path of vertices  $s, y, v, \bar{x}, x, \bar{w}, w, \bar{u}, t$ , and so we set  $\psi(y) = u$ ).

Since each vertex  $\bar{u}$ , with  $u \in B \setminus F$ , can receive at most  $L_u$  units of flow, clearly  $\psi$  respects the capacities. Moreover, since there are at most  $\alpha - 1$  elements in  $B \cap F$ , each unit of flow leaving a vertex  $v$  in  $F$  can traverse at most  $\alpha - 1$  reverse arcs in  $H$  (without creating a circle), so it can traverse at most  $\alpha$  arcs from  $F$  before reaching an element  $\bar{u}$  in  $\bar{B} \setminus \bar{F}$ . Therefore,  $d_G(v, u) \leq 6\alpha$ .

It follows that for every  $y \in \phi^{-1}(F)$

$$d_G(y, \psi(y)) \leq d_G(y, \phi(y)) + d_G(\phi(y), \psi(y)) \leq \beta + 6\alpha.$$

Now we can define a conservative reassignment  $\phi_F$ :

$$\phi_F(v) = \begin{cases} \phi(v) & \text{if } \phi(v) \notin F, \\ \psi(v) & \text{otherwise.} \end{cases}$$

We argue that  $\phi_F$  is a valid distance- $(\beta + 6\alpha)$  conservative reassignment. Let  $u$  be a center opened by the algorithm (that is,  $u \in S \cup B$ ). If  $u \in F$ , then  $\phi_F(v) = \emptyset$ ; otherwise,  $\phi_F^{-1}(u) = \phi^{-1}(u)$  if  $u \in S$ , or  $\phi_F^{-1}(u) = \psi^{-1}(u)$  if  $u \in B$ . Since  $\phi$  and  $\psi$  do not exceed the capacities of the centers to which they assign clients, neither does  $\phi_F$ . Finally, consider a vertex  $y$  in  $V$ . If  $\phi(y) \notin F$ , then  $\phi_F(y) = \phi(y)$  and  $d_G(y, \phi_F(y)) = d_G(y, \phi(y)) \leq \beta$ . If  $\phi(y) \in F$ , then  $d_G(y, \phi_F(y)) = d_G(y, \psi(y)) \leq \beta + 6\alpha$ .  $\square$

Using the approximation for the capacitated  $\kappa$ -center by An et al. [10], we obtain the following.

**Corollary 2** *Algorithm 2 using the algorithm by An et al. [10] for the capacitated  $\kappa$ -center is a  $(9 + 6\alpha)$ -approximation for the capacitated conservative  $\alpha$ -fault-tolerant  $\kappa$ -center with fixed  $\alpha$ .*

## 5 Capacitated Fault-Tolerant $k$ -Center

### 5.1 An Initial LP Formulation

Recall that we are given an unweighted connected graph, and the objective is to decide whether there is a distance-1 solution (see Sect. 2). As in [9, 10], we use an integer LP that formulates the problem. If, after relaxing the integrality constraints, the LP is infeasible, then we know that there is no distance-1 solution; otherwise, we round the solution and obtain an approximate solution.

In the natural formulation for the capacitated  $\kappa$ -center, we have opening variables  $y_u$  for each vertex  $u$ , representing the choice of  $u$  as a center, and assignment variables  $x_{uv}$  representing that vertex  $v$  is assigned to center  $u$ . In the case of the fault-tolerant  $\kappa$ -center, for each failure scenario, that is, for each possible set  $F \subseteq V$  of centers that may fail, with  $|F| \leq \alpha$ , we must have a different assignment from vertices to non-faulty centers opened by  $y$ . One possibility to formulate the fault-tolerant variant is having different assignment variables for each  $F$ . To simplify the formulation, rather than creating a different set of assignment variables for each failure scenario, we use an equivalent formulation based on Hall's condition, which is a necessary and sufficient condition for a bipartite graph to have a perfect matching [16]. The integer linear program, denoted by  $ILP_{k,\alpha}(G)$ , is the following:

$$\begin{aligned} \sum_{u \in V} y_u &= k \\ |U| &\leq \sum_{u \in N_G(U) \setminus F} y_u L_u & \forall U \subseteq V, F \subseteq V : |F| = \alpha \\ y_u &\in \{0, 1\} & \forall u \in V. \end{aligned}$$

We remark that  $ILP_{k,\alpha}(G)$  formulates the capacitated  $\alpha$ -fault-tolerant  $\kappa$ -center. The first constraint guarantees that exactly  $k$  centers are opened, and the second set of constraints guarantees that, for each failure scenario, there is a feasible assignment from clients to opened centers that did not fail. Indeed, notice that, for a fixed  $F$ , the existence of such an assignment is equivalent to the existence of a matching on the bipartite graph formed by clients and open units of capacity that matches all clients. Hall's result, together with the second set of constraints of  $ILP_{k,\alpha}(G)$ , assures the existence of such a matching, and thus of such an assignment. Notice that we omit failure scenarios  $F$  with  $|F| < \alpha$  in the formulation, as the corresponding inequalities are implied using inequalities for  $F'$  with  $|F'| = \alpha$  and  $F \subseteq F'$ .

### 5.1.1 Integrality Gap

As a first attempt, one can relax  $ILP_{k,\alpha}(G)$  directly. When the integrality constraints are relaxed, however, the total value of fractional openings of the centers in  $F$  could be strictly less than  $\alpha$ , that is,  $y(F) < \alpha$ . Thus the considered constraints are weaker than desired. Indeed, consider the following example. Let  $C_n$  be a cycle on  $n$  vertices, for  $n = s^2$  where  $s$  is a positive even integer, and let  $G_n$  be the graph obtained from  $C_n$  by adding edges between every two vertices at distance at most  $s$  in  $C_n$ . Note that any pair of antipodes in  $C_n$  are at distance  $s/2$  in  $G_n$ . If  $L_u = n$  for every  $u$  in  $G_n$ ,  $k = s$ , and  $\alpha = k - 1$ , then the cost of any solution for this instance is  $s/2$ , as for any set of  $k$  centers in  $G_n$ , all but one center might fail. Now, let  $y$  be the vector with  $y_u = 1/s$  for every  $u$ . We claim that  $y$  is feasible for the relaxation of  $ILP_{k,\alpha}(G_n)$ . Indeed, first notice that  $\sum_{u \in V} y_u = s = k$ . Also, since every vertex has  $2s$  neighbors, for any set of centers  $F$  of size  $\alpha = k - 1 = s - 1$ , the second set of constraints is satisfied, because either  $U = \emptyset$ , and the constraint is trivially satisfied, or the right side is at least  $n$ , and the left side is at most  $n$ . So  $y$  is feasible, and thus the lower bound obtained from the relaxation of  $ILP_{k,\alpha}(G_n)$  may be arbitrarily small when compared to an optimal solution, that is, the minimization problem obtained from  $ILP_{k,\alpha}(G_n)$  has unbounded integrality gap.

## 5.2 Dealing With the Integrality Gap

Suppose that we knew a subset  $B$  of the centers of an optimal solution that might fail. Then we could set  $y_u = 1$  for each  $u$  in  $B$ , that is, we force the LP to open  $u$ . This would avoid the problem in the example with unbounded integrality gap whenever the failure scenario is  $F \subseteq B$ , as in such a case we would have  $y(F) = |F|$ . Since we do not know how to obtain a subset  $B$  of centers of an optimal solution, and a failure scenario  $F$  might contain centers not in  $B$ , we aim at two more relaxed goals:

- (G1) Based on the structure of  $G$ , we determine approximate locations of centers in an optimal solution. This allow us to select a subset of centers  $B$  that are close to distinct centers of such an optimal solution.
- (G2) We consider first the case where only centers in  $B$  might fail, and carefully choose  $B$  so that this case comprises the worst scenario.

To achieve these goals, we will make use of a standard clustering technique. We partition the graph so that the elements of each part are close to some centers in an optimal solution. Locally, the worst-case scenario corresponds to the failure of the highest capacitated centers in a cluster. The clustering and the selection of pre-opened centers are described precisely in the following.

### 5.2.1 Clustering

Clustering has been used by several algorithms for the  $\kappa$ -center problem, for both the capacitated [7,8] and fault-tolerant cases [12–15]. We use the construction introduced by Khuller and Sussmann [8]. In their algorithm, one first repeatedly selects a new

center at distance exactly 3 from the set of previous selected centers, and then attaches each vertex to its nearest center (breaking ties arbitrarily). In the end of this process, each vertex is either a center, or is at distance at most 2 from the attached center. The relevant result is replicated in next lemma.

**Lemma 4** [8] *Given a connected graph  $G = (V, E)$ , one can obtain a set of midpoints  $\Gamma \subseteq V$ , and a partition of  $V$  into sets  $\{C_v\}_{v \in \Gamma}$ , such that*

- *there exists a rooted tree  $T$  on  $\Gamma$ , with  $d_G(u, v) = 3$  for every edge  $(u, v)$  of  $T$ ;*
- *$N_G(v) \subseteq C_v$  for every  $v$  in  $\Gamma$ ; and*
- *$d_G(u, v) \leq 2$  for every  $v$  in  $\Gamma$  and every  $u$  in  $C_v$ .*

### 5.2.2 Selecting Pre-opened Centers

We apply Lemma 4 and obtain a clustering of  $V$ . Let  $v$  in  $\Gamma$  be a cluster midpoint, and suppose there exists a distance-1 solution for  $G$ . Since up to  $\alpha$  centers in this solution may fail, there must be at least  $\alpha + 1$  centers in  $N(v)$ , as otherwise there would be a failure scenario for which  $v$  would not have a surviving center in its neighborhood. Thus, the elements of  $C_v$  are within distance 3 from at least  $\alpha + 1$  centers in  $C_v$ . Moreover, since sets  $N(v)$  are disjoint for  $v$  in  $\Gamma$ , there are at least  $\alpha + 1$  centers per cluster in any distance-1 solution. If  $|N(v)| \leq \alpha$  for a vertex  $v$ , then  $ILP_{k,\alpha}(G)$  is trivially unfeasible, and we obtain a certificate that the input graph is a no instance. In the following, we assume that  $|N(v)| \geq \alpha + 1$  for every vertex  $v$ .

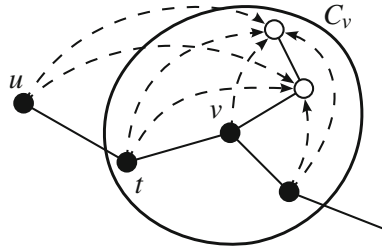
To achieve (G1), we may select, for each cluster, any subset of up to  $\alpha + 1$  vertices in the cluster. To achieve (G2), we reason on the total capacity that may become unavailable when failure occurs. For each cluster, the largest amount of capacity that can be discounted in a given scenario does not exceed the accumulated capacity of the  $\alpha$  most capacitated vertices in the cluster. Thus, we select these vertices as set  $B$ .

Formally, for each  $v$  in  $\Gamma$ , let  $B_v \subseteq C_v$  be a set of  $\alpha$  elements of  $C_v$  with largest capacities. This is the set of pre-opened centers for cluster  $C_v$ . The set of all pre-opened centers is defined as

$$B = \bigcup_{v \in \Gamma} B_v.$$

### 5.3 Modifying the LP Formulation

We pre-open the elements of  $B$  by adding to  $ILP_{k,\alpha}(G)$  the constraint  $y_u = 1$ , for every  $u \in B$ . When we establish a partial solution in advance, we may turn the original linear formulation infeasible, since it is possible that no distance-1 solution opens the elements of  $B$ . However, since in any distance-1 solution there are at least  $\alpha$  centers in  $N_G(v)$  for a given cluster midpoint  $v$ , each such center is within distance 3 to a distinct element of  $B$  of non-smaller capacity. Thus, we can convert a distance-1 solution into a distance-4 solution by reassigning clients to elements of  $B$ , while preserving most of the structure in the original LP.



**Fig. 2** Dashed lines represent arcs added to the graph  $G$  to obtain  $G'$ , and solid lines represent duplicated arcs in opposite directions. Empty circles represent  $B_v$

### 5.3.1 Fixing Feasibility

To obtain a useful LP relaxation, while pre-opening the set  $B$  of centers, we modify the supporting graph  $G$ . For each cluster  $C_v$ , we augment  $G$  with edges connecting each client that could be potentially served by centers in  $C_v$  to each vertex in the set  $B_v$ . Precisely, we define the directed graph  $G' = G'(G, \{C_v\}_{v \in \Gamma}) = (V, E')$ , where  $E'$  is the set of arcs  $(u, w)$  such that  $\{u, w\} \in E$ , or there exist  $v \in \Gamma$  and  $t \in N_G(v)$  such that  $\{u, t\} \in E$  and  $w \in B_v$  (see Fig. 2). We remark that a directed graph is used, because we want to allow for a reassignment of a client from an arbitrary center in the cluster to a center in  $B$ , but not the other way around.

### 5.3.2 A New Formulation

In the new formulation, we consider only scenarios  $F \subseteq B$ . Thus, in a feasible solution  $y$ , we will have  $y(F) = |F|$  for each scenario  $F$ . Also, for each cluster midpoint, we want the total value of fractional openings of non-faulty centers in its neighborhood to be at least one. For the integer program  $ILP_{k,\alpha}(G)$ , this was implicit by the constraints, but when  $y$  is not integral, there might be high capacity centers that satisfy the local demand with less than one open unit. Therefore, we have an additional constraint for each cluster midpoint  $v$  to ensure that there is one unit of (fractional) opening in  $N_G(v)$  excluding any opening coming from  $B$ . We obtain a new linear program, denoted by  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$ .

$$\begin{aligned} \sum_{u \in V} y_u &= k \\ |U| &\leq \sum_{u \in N_{G'}(U) \setminus F} y_u L_u & \forall U \subseteq V, F \subseteq B : |F| = \alpha \\ 1 &\leq \sum_{u \in N_G(v) \setminus B} y_u & \forall v \in \Gamma \\ y_u &= 1 & \forall u \in B \\ 0 &\leq y_u \leq 1 & \forall u \in V. \end{aligned}$$

Notice that, contrary to  $ILP_{k,\alpha}(G)$ , program  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$  depends on the obtained clustering. The following lemma states that  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$  is a “relaxation” of  $ILP_{k,\alpha}(G)$ , that is, if  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$  is infeasible, then we obtain a certificate that no distance-1 solution for  $G$  exists.



**Lemma 5** *If  $ILP_{k,\alpha}(G)$  is feasible, then  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$  is feasible.*

*Proof* Suppose that  $ILP_{k,\alpha}(G)$  is feasible. Let  $y$  be a feasible solution for  $ILP_{k,\alpha}(G)$ , and let  $R$  be the set of centers corresponding to  $y$ .

First, we define an injection  $\beta$  from  $R$  into  $R \cup B$  that covers  $B$ . We begin by defining  $\beta(w) = w$  for each  $w \in R \cap B$ . Recall that  $\Gamma$  is the set of midpoints. Now, for each  $v$  in  $\Gamma$ , let  $u_1, \dots, u_t$  be the elements of  $B_v \setminus R$  in non-increasing order of capacity. Analogously, let  $w_1, \dots, w_t, \dots$  be the elements of  $(R \cap N_G(v)) \setminus B$  in non-increasing order of capacity (recall that each  $N_G(v)$  has at least  $\alpha + 1$  centers in an optimal solution  $R$ , and thus  $|(R \cap N_G(v)) \setminus B| > |B_v \setminus R|$ ). For each  $i$  with  $1 \leq i \leq t$ , we define  $\beta(w_i) = u_i$ . Finally, for each  $w$  in  $R$  whose  $\beta(w)$  is not defined yet, let  $\beta(w) = w$ . Notice that  $\beta$  covers  $B$ ,  $L_w \leq L_{\beta(w)}$  for every  $w$  in  $R$ , and the inverse function  $\beta^{-1}$  is well-defined on the image of  $\beta$ .

Consider a set  $U \subseteq V$ . We claim that  $\beta(N_G(U)) \subseteq N_{G'}(U)$ . Indeed, let  $u \in U$ , and  $t \in N_G(\{u\})$ . Then, if  $\beta(t) = t$ , we have  $\beta(t) \in N_G(U) \subseteq N_{G'}(U)$ . Otherwise,  $t$  must be a neighbor of some midpoint  $v$ , and  $\beta(t) \in B_v$ . In this case we know that  $(u, \beta(t))$  is an edge of  $G'$ , and thus  $\beta(t) \in N_{G'}(U)$ .

Let  $R' = \beta(R)$ , and let  $y'$  be the characteristic vector of  $R'$ . We claim that  $y'$  is a feasible solution for  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$ . Let  $U \subseteq V$  and  $F \subseteq B$  with  $|F| = \alpha$ . From the feasibility of  $y$  for  $ILP_{k,\alpha}(G)$ , and as  $|\beta^{-1}(F)| = |F| = \alpha$ , we have

$$\begin{aligned} |U| &\leq \sum_{u \in N_G(U) \setminus \beta^{-1}(F)} y_u L_u = \sum_{u \in (N_G(U) \setminus \beta^{-1}(F)) \cap R} L_u \\ &= \sum_{u \in (N_G(U) \cap R) \setminus \beta^{-1}(F)} L_u \leq \sum_{u \in \beta((N_G(U) \cap R) \setminus \beta^{-1}(F))} L_u \\ &= \sum_{u \in \beta(N_G(U) \cap R) \setminus F} y'_u L_u \leq \sum_{u \in N_{G'}(U) \setminus F} y'_u L_u, \end{aligned}$$

where the last inequality holds since  $\beta(N_G(U)) \subseteq N_{G'}(U)$ . The verification that the other constraints also hold for  $y'$  is straightforward.  $\square$

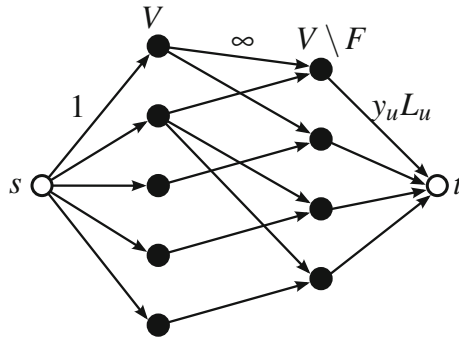
Though  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$  has an exponential number of constraints, the following lemma shows that it has a polynomial-time separation oracle [17, Chap. 14].

**Lemma 6** *For fixed  $\alpha$ , there is an algorithm that, in polynomial time, decides whether a vector  $y$  is feasible for  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$ . If  $y$  is not feasible, the algorithm also outputs a constraint of  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$  that is violated by  $y$ .*

*Proof* We concentrate on the second set of constraints, as there are polynomially many constraints of the other types. Notice that the number of distinct scenarios  $F$  is  $O(|V|^\alpha)$ , which is polynomial since  $\alpha$  is constant. Fix a failure scenario  $F$  and suppose that we can solve the following problem:

$$\min_{U \subseteq V} \sum_{u \in N_{G'}(U) \setminus F} y_u L_u - |U|. \quad (1)$$

If this value is non-negative, then all constraints in the second set for this scenario  $F$  are satisfied, otherwise there is a subset  $U^*$  of  $V$  for which the constraint is violated,



**Fig. 3** Flow network with source  $s$  and target  $t$ . There is a unit-capacity arc from  $s$  to each  $v \in V$ , and an arc with infinite capacity from  $v$  to each neighbor  $u \in N_{G'}(U) \setminus F$ . Moreover, for each  $u \in V \setminus F$ , there is an arc to  $t$  with capacity  $y_u L_u$

and we are done. We can rewrite the minimization problem above as the following integer linear program on binary variables  $a_u$  for  $u$  in  $V$ , and  $b_u$  for  $u$  in  $V \setminus F$ :

$$\begin{aligned} \min \quad & \sum_{u \in V \setminus F} b_u (y_u L_u) + \sum_{u \in V} a_u - |V| \\ \text{s.t.} \quad & a_u + b_v \geq 1 \quad \forall (u, v) \in E(G') \\ & a_u, b_v \in \{0, 1\} \quad \forall u \in V, v \in V \setminus F. \end{aligned}$$

Variable  $a_u$  indicates that  $u$  is not in  $U$ , and variable  $b_v$  indicates that there exists some  $u$  in the adjacency list of  $v$  that is in  $U$  (that is,  $a_u = 0$ ). The corresponding matrix for this problem is totally unimodular, so the relaxation has an integral optimal solution, which can be found in polynomial time [17, Example 1, p. 273]. Notice that this problem (excluding the constant  $-|V|$  in the objective function) corresponds to the min-cut formulation for the network flow problem depicted in Fig. 3, so it suffices to run any max-flow min-cut algorithm.  $\square$

**Corollary 3** For fixed  $\alpha$ ,  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$  can be solved in polynomial time.

If, for an arbitrary  $\alpha$ , there were a polynomial-time algorithm for finding a set  $F$  with  $|F| = \alpha$  that minimizes the value of (1), then a stronger version of Corollary 3 without the restriction on  $\alpha$  being fixed would hold. In Sect. 8, we give evidence that this algorithm only exists if  $P = NP$ .

## 5.4 Distance- $r$ Transfers

Given a solution  $y$  for  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$ , the problem of finding  $k$  centers to serve all clients is now reduced to rounding vector  $y$  so that exactly  $k$  vertices are integrally open. Since the total fractional opening of  $y$  is  $k$ , one might consider “moving” the fractional opening from one vertex to another so that the opening of some vertices becomes zero, while the opening of  $k$  vertices become one. This idea motivated the distance- $r$  transfers, introduced by An et al. [10], which we adapt to the fault-tolerant context.

In a distance- $r$  transfer, the fractional opening of vertices is moved to vertices within distance at most  $r$ . This guarantees that, after performing transfer operations, the cost of the solution grows in a controlled way. To ensure that the capacity constraints are not violated, one might consider only transferring fractional opening from a low capacitated vertex to higher capacitated vertices, so that the “local capacity” does not decrease. Here, we use a slightly more general definition than the original one to comprise our requirements, as we might need to ensure that the opening of certain vertices is never transferred (that is the case for vertices in  $B$ ), and that transfers follow certain paths.

**Definition 2** Let  $V$  be a set of vertices,  $W$  be a subset of  $V$ ,  $H$  be a graph on  $W$ , and  $L : V \rightarrow \mathbb{R}_{\geq 0}$  be a capacity function on  $V$ . A vector  $y' : V \rightarrow \mathbb{R}_{\geq 0}$  is an  $H$ -restricted distance- $r$  transfer of a vector  $y : V \rightarrow \mathbb{R}_{\geq 0}$  if

- (a)  $\sum_{v \in V} y'_v = \sum_{v \in V} y_v$ ;
- (b)  $\sum_{v \in N_H^r(U)} L_v y'_v \geq \sum_{v \in U} L_v y_v$  for every  $U \subseteq W$ ; and
- (c)  $y'_v = y_v$  for every  $v \in V \setminus W$ .

If  $y'$  is the characteristic vector of a set  $R \subseteq V$ , we will say that  $R$  is an *integral  $H$ -restricted distance- $r$  transfer* of  $y$ . If the graph  $H$  is clear from the context, then we might simply say that  $y'$  is a *distance- $r$  transfer* of  $y$ .

An et al. [10] reduced the rounding of an arbitrary graph to the case in which the graph is a tree that satisfies certain properties. They showed that such trees have integral distance-2 transfers. This is formalized in the following.

**Lemma 7** ([10]) Let  $T = (W, E)$  be a tree, and  $y : W \rightarrow [0, 1]$  be a vector such that  $\sum_{u \in W} y_u$  is an integer, and  $y_v = 1$  for every internal node  $v$  of  $T$ . One can find in polynomial time an integral ( $T$ -restricted) distance-2 transfer of  $y$ .

For a given solution  $y$  for  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$  and any failure scenario  $F \subseteq B$ , the LP implicitly defines an assignment of clients to non-faulty (fractionally opened) centers at distance 1 in  $G'$ . Suppose some portion of the opening  $y_v$  of  $v \notin B$  is transferred to some other vertex  $v'$  at distance  $r$  in  $G$ . If a client  $u$  is initially served by  $v$ , then the assignment can be transferred to  $v'$  as well, so that  $u$  will be (fractionally) assigned to centers at distance at most  $r + 1$  in  $G$ , as  $(u, v) \in E(G)$ . If client  $u$  was initially served by some  $v \in B$ , then this assignment may be left unchanged, as no opening of  $v$  is transferred; in this case, however, we might have  $(u, v) \notin E(G)$ , and so edge  $(u, v)$  in  $G'$  may correspond to a path of length 4 in  $G$ . The worst case of the obtained assignment happens when the distance is the maximum between  $r + 1$  and 4.

## 5.5 The Algorithm

Our algorithm consists of two parts. In the first, we round a fractional solution  $y$  of  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$ , and obtain a set  $R$  of  $k$  centers. In the second part, for each failure scenario  $F \subseteq R$  with  $|F| \leq \alpha$ , we have to obtain an assignment from  $V$  to  $R \setminus F$ .

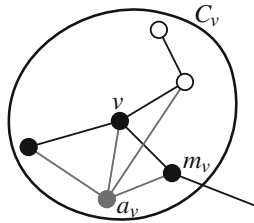
### 5.5.1 Rounding

Since we have pre-opened centers in  $B$ , we round only the values associated with vertices in  $V \setminus B$ . This phase is based on the algorithm of An et al. [10] for the capacitated (non-fault-tolerant)  $\kappa$ -center. The main difference is that we do not allow transfers from or to vertices in the set  $B$ . The algorithm reduces the problem of rounding the solution for an arbitrary graph to the particular case of the problem whose instances are trees. There are three consecutive transfers. In the first step, we concentrate one unit of opening in an auxiliary vertex that is added at the same location as the cluster midpoint. In the second step, we create a tree instance using the auxiliary vertices as internal nodes, and obtain an integral transfer using Lemma 7. In the last step, the opening of auxiliary vertices is transferred back to vertices of the original graph. A detailed description is presented in the following:

- Step 1. For each cluster  $C_v$ , choose an element  $m_v$  in the neighborhood of the midpoint  $v$  that is not pre-opened, and has the largest capacity, that is,  $m_v = \arg \max_{u \in N_G(v) \setminus B} L_u$ . Create an auxiliary vertex  $a_v$  at the same location as  $v$  (add an edge to  $a_v$  from each element of  $N_G(v)$  as in Fig. 4), with capacity  $L_{a_v} = L_{m_v}$ , and initial opening  $y_{a_v} = 0$ . Next, aggregate one unit of opening to  $a_v$  by transferring fractional openings from  $N_G(v) \setminus B$  to  $a_v$ . This can be done as  $\sum_{u \in N_G(v) \setminus B} y_u \geq 1$ . The transfer proceeds as follows: for each  $u$  in  $N_G(v) \setminus B$ , decrease  $y_u$ , while increasing  $y_{a_v}$  until  $y_u$  becomes 0. The process is interrupted once  $y_{a_v}$  reaches 1. The result is a distance-1 transfer  $y^{(1)}$ . The first vertex to have its fractional opening transferred is  $m_v$ , so that, at the end of this step,  $y_{m_v}^{(1)} = 0$ .
- Step 2. Construct a tree  $T$  from the clustering tree by replacing each midpoint  $v$  with  $a_v$  for every  $v$  in  $\Gamma$ . Next, for each cluster  $C_v$ , select every vertex  $u$  in  $C_v$  such that  $u \neq m_v$  and  $u \notin B$  and add a leaf corresponding to  $u$ , connected to  $a_v$ . Finally, apply Lemma 7 to obtain an integral  $T$ -restricted distance-2 transfer  $y^{(2)}$  (starting with  $y^{(1)}$  restricted to the vertices of  $T$ , and then extending the vector that results from the lemma to include the other vertices with values unchanged). Notice that  $d_G(w_1, w_2) = 3$  for each edge  $(w_1, w_2)$  of  $T$  if both  $w_1$  and  $w_2$  are internal nodes; and  $d_G(w_1, w_2) \leq 2$  if either  $w_1$  or  $w_2$  is a leaf. Hence,  $y^{(2)}$  can be interpreted as a distance- $(2 \cdot 3)$  transfer of  $y^{(1)}$  (on the basis graph).
- Step 3. For each cluster  $C_v$ , transfer the opening of the auxiliary vertex  $a_v$  back to the original vertex  $m_v$ . This is possible since  $m_v$  is not part of  $T$ , and thus  $y_{m_v}^{(2)} = 0$ . Obtain a final integral distance-1 transfer  $y^{(3)}$ . Open the set of vertices  $R \subseteq V$  that corresponds to the characteristic vector  $y^{(3)}$ .

### 5.5.2 Assignment

After opening centers  $R$ , up to  $\alpha$  failures might occur. Our algorithm must provide a valid assignment for each failure scenario  $F \subseteq R$ .



**Fig. 4** Auxiliary vertex  $a_v$  is adjacent to each vertex in  $N_G(v)$ . The white vertices represent set  $B_v$

First, we examine the case that  $F$  is a subset of  $B$ . In this case,  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$  assures the existence of an assignment from  $V$  to a set of fractionally opened centers that does not intersect  $F$ . Since the rounding algorithm obtains an integral  $G$ -restricted distance-8 transfer (by adding up the three consecutive transfers), this will lead to a distance-9 solution that does not assign to any element of  $F$ .

For the case that  $F$  is not a subset of  $B$ , we may not rely on the existence of a fractional assignment obtained from the LP. Instead, we will show that, for each  $F$ , there exists a corresponding  $F' \subseteq B$ , and that a distance-9 solution for failure scenario  $F'$  can be transformed into a distance-10 solution for failure scenario  $F$ . More precisely, we will show that each element  $u$  assigned to a center  $v \in F \setminus F'$  in the former solution may be reassigned to a distinct element  $v' \in F' \setminus F$  in the latter solution, such that  $v$  and  $v'$  are in the same cluster, and  $L_{v'} \geq L_v$ .

A naive analysis of the preceding strategy would yield a 13-approximation, as the distance between  $v$  and  $v'$  might be 4, and thus  $d(u, v') \leq d(u, v) + d(v, v') \leq 9 + 4$ . To obtain a more refined analysis, we will bound the distance between  $u$  and the midpoint associated to  $v$ . More precisely, denote by  $\delta(v)$  the midpoint of the cluster that contains  $v$ . We obtain the following lemma.

**Lemma 8** Consider  $F \subseteq B$  with  $|F| = \alpha$  and let  $R$  be the integral transfer obtained from  $y$  by the rounding algorithm above. One can find an assignment  $\phi : V \rightarrow R \setminus F$  such that  $d_G(u, \delta(\phi(u))) \leq 8$  and  $d_G(u, \phi(u)) \leq 9$  for each  $u$  in  $V$ .

*Proof* Let  $\bar{V} = V \cup \{a_v : v \in \Gamma\}$  be the union of the vertices of  $G$  and the auxiliary vertices, and  $\bar{G}$  be the graph obtained after we add the auxiliary vertices to  $G$ . Fix a subset  $U \subseteq V$ .

Recall that the rounding algorithm considers an initial feasible solution  $y = y^{(0)}$ , and obtains consecutive transfers  $y^{(1)}, y^{(2)}, y^{(3)}$ . In the following, for each transfer  $y^{(i)}$ , for  $0 \leq i \leq 2$ , and each  $U \subseteq V$ , we will consider a set  $X = X^{(i)}(U)$  of nearby vertices whose total installed capacity, excluding faulty elements, exceeds  $|U|$ . That is, we want to obtain  $X$  such that the value  $ic^{(i)}(X) := \sum_{u \in X \setminus F} y_u^{(i)} L_u \geq |U|$ . Initially, before any transfer is performed, we have that  $y^{(0)} = y$  and, by the constraints of  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$ , we have that  $|U| \leq \sum_{u \in N_{G'}(U) \setminus F} y_u L_u$ , so we set  $X^{(0)}(U) = N_{G'}(U)$ .

In the first step, we have a distance-1 transfer. Notice that  $N_{G'}(U) \setminus B = N_G(U) \setminus B$ . Also, recall that the transfer is restricted to vertices in  $\bar{V} \setminus B$ . We obtain

$$\begin{aligned}
 |U| &\leq ic^{(0)}(X^{(0)}(U)) = ic^{(0)}((N_{G'}(U) \cap B) \cup (N_G(U) \setminus B)) \\
 &\leq ic^{(1)}((N_{G'}(U) \cap B) \cup (N_G^2(U) \setminus B)) \\
 &= ic^{(1)}((N_{G'}(U) \cap B) \cup (N_G^2(U) \cap V(T))).
 \end{aligned}$$

The last equality holds because, if  $u \in N_G^2(U) \setminus B$  but  $u \notin V(T)$ , then  $u = m_v$  for some  $v \in \Gamma$ , thus  $y_u^{(1)} = 0$ . Hence we set  $X^{(1)}(U) = (N_{G'}(U) \cap B) \cup (N_G^2(U) \cap V(T))$ .

In the second step, we have an integral  $T$ -restricted distance-2 transfer of  $y^{(1)}$ . Once again, since  $T$  does not include vertices of  $B$ , we obtain

$$\begin{aligned}
 ic^{(1)}(X^{(1)}(U)) &= ic^{(1)}((N_{G'}(U) \cap B) \cup (N_G^2(U) \cap V(T))) \\
 &\leq ic^{(2)}((N_{G'}(U) \cap B) \cup N_T^2(N_G^2(U) \cap V(T))).
 \end{aligned}$$

We set  $X^{(2)}(U) = (N_{G'}(U) \cap B) \cup N_T^2(N_G^2(U) \cap V(T))$ .

Let  $\bar{R} \subseteq \bar{V}$  be the set corresponding to vector  $y^{(2)}$ . First consider a bipartite graph  $H = (V \cup \bar{R}, D)$ , where elements  $v \in V \cap \bar{R}$  appear duplicated: the original in the first side of the bipartition, and a copy in the second side. There is an edge  $\{u, v\}$  in  $D$  if  $v \in X^{(2)}(\{u\}) \setminus F$ . Modify  $H$  by replacing each vertex  $v$  in  $\bar{R}$  by  $L_v$  copies. Notice that now, for each  $U \subseteq V$ , we have  $|N_H(U) \setminus U| = |\bigcup_{u \in U} (N_H(\{u\}) \setminus \{u\})| = ic^{(2)}(\bigcup_{u \in U} X^{(2)}(\{u\})) = ic^{(2)}(X^{(2)}(U)) \geq |U|$ . This is exactly Hall's condition for the existence of a matching in  $H$  covering  $V$ . We obtain such a matching and a corresponding assignment  $\phi' : V \rightarrow \bar{R} \setminus F$ .

Remember that  $\delta(u)$  is the midpoint of the cluster that contains  $u \in V$ . For an auxiliary vertex  $a_v$  with  $v \in \Gamma$ , we let  $\delta(a_v) = v$ . Now, for every vertex  $u$  in  $V$ , we show that the distance from  $u$  to both  $\phi'(u)$  and  $\delta(\phi'(u))$  is bounded by 8. Recall that  $\phi'(u) \in X^{(2)}(\{u\})$ . We have two cases. First, suppose that  $\phi'(u) \in N_{G'}(\{u\}) \cap B$ . Since we have that  $d_G(u, \phi'(u)) \leq 4$ , by the construction of  $G'$ , we obtain that  $d_G(u, \delta(\phi'(u))) \leq 6$ , and we are done. Now, assume that  $\phi'(u) \in N_T^2(N_G^2(\{u\}) \cap V(T))$ . In this case, there must be some  $v$  in  $N_G^2(\{u\}) \cap V(T)$  and a shortest path  $\rho$  connecting  $v$  to  $\phi'(u)$  in  $T$ . We consider two possibilities. If the length of  $\rho$  is 1, then  $\rho = (v, \phi'(u))$ , and we deduce that  $d_{\bar{G}}(u, \delta(\phi'(u))) \leq d_{\bar{G}}(u, v) + d_{\bar{G}}(v, \phi'(u)) + d_{\bar{G}}(\phi'(u), \delta(\phi'(u))) \leq 2 + 3 + 2 = 7$ . If the length of  $\rho$  is 2, then there exists  $w$  such that  $\rho = (v, w, \phi'(u))$ , and we get that  $d_{\bar{G}}(u, \phi'(u)) \leq d_{\bar{G}}(u, v) + d_{\bar{G}}(v, w) + d_{\bar{G}}(w, \phi'(u)) \leq 2 + 3 + 3 = 8$ . Let  $z = \delta(\phi'(u))$  be the midpoint corresponding to  $\phi'(u)$ . If  $\phi'(u)$  is an internal node of  $T$ , then it must be the auxiliary vertex  $a_z$ , and thus  $d_{\bar{G}}(u, \delta(\phi'(u))) = d_{\bar{G}}(u, z) \leq d_{\bar{G}}(u, \phi'(u)) \leq 8$ , because  $z$  and  $\phi'(u) = a_z$  are at the same location. Otherwise,  $w$  must be an internal node and  $\phi'(u)$  a leaf of  $w$ . Hence  $w = a_z, \delta(\phi'(u)) = z$ , and therefore  $d_{\bar{G}}(u, \delta(\phi'(u))) = d_{\bar{G}}(u, z) \leq d_{\bar{G}}(u, w) \leq d_{\bar{G}}(u, v) + d_{\bar{G}}(v, w) \leq 2 + 3 = 5$ .

To obtain a final assignment  $\phi : V \rightarrow R \setminus F$ , we reassign each vertex  $u$  assigned to an auxiliary vertex  $a_v$ , to the vertex  $m_v$ , that is, for each  $u$  in  $V$ , if  $\phi'(u) = a_v$  for some  $v$  in  $\Gamma$ , then set  $\phi(u) = m_v$ ; otherwise, set  $\phi(u) = \phi'(u)$ . Since  $d(u, \delta(\phi(u))) = d(u, \delta(\phi'(u)))$  and  $d(u, \phi(u)) \leq d(u, \phi'(u)) + 1$ , we conclude that  $d(u, \delta(\phi(u))) \leq 8$  and  $d(u, \phi(u)) \leq 9$ .  $\square$

Now we may obtain the approximation factor.

**Theorem 3** *There exists a 10-approximation for the capacitated  $\alpha$ -fault-tolerant  $\kappa$ -center with fixed  $\alpha$ .*

*Proof* Consider a failure scenario  $F \subseteq R$  with  $|F| = \alpha$ . For each cluster  $C_v$ , let  $F_v$  be the set of centers that failed in cluster  $C_v$ . Also, let  $F'_v$  be the set of the  $|F_v|$  most capacitated centers in  $B_v$ , and  $F' = \bigcup_{v \in \Gamma} F'_v$ . We use Lemma 8, and obtain an assignment  $\phi' : V \rightarrow R \setminus F'$ . In the following, we define an assignment  $\phi : V \rightarrow R \setminus F$  that respects the capacities, and such that  $d_G(w, \phi(w)) \leq 10$  for  $w \in V$ .

First, let  $w \in V$  be such that  $\phi'(w) \in R \setminus (F \cup F')$ . In this case, set  $\phi(w) = \phi'(w)$ , and notice that  $d_G(w, \phi(w)) = d_G(w, \phi'(w)) \leq 9$ , by Lemma 8. Hence, since  $\phi'$  respects the capacities, so does  $\phi$  for centers in  $R \setminus (F \cup F')$ .

Now consider vertices  $w \in V$  such that  $\phi'(w) \in F$ . For each  $v$  in  $\Gamma$ , obtain an ordering  $\{u_1, \dots, u_t\}$  of the vertices in  $F'_v \setminus F_v$  and an ordering  $\{v_1, \dots, v_t\}$  of the vertices in  $F_v \setminus F'_v$ . For every  $w$  that is assigned to  $v_i$ , for some  $1 \leq i \leq t$ , reassign it to  $u_i$ , that is, for every  $w$  such that  $\phi'(w) = v_i$ , set  $\phi(w) = u_i$ . Notice that this leads to a valid assignment  $\phi$ , since  $L_{u_i} \geq L_{v_i}$  for every  $1 \leq i \leq t$ . Since  $u_i$  and  $v_i$  are in the same cluster,  $d_G(\delta(v_i), u_i) \leq 2$ , and thus  $d_G(w, \phi(w)) \leq d_G(w, \delta(\phi'(w))) + d_G(\delta(\phi'(w)), u_i) \leq 8 + 2 = 10$ .  $\square$

## 6 $\{0, L\}$ -Capacitated Fault-Tolerant $k$ -Center

For a given  $L$ , the  $\{0, L\}$ -capacitated fault-tolerant  $\kappa$ -center is the particular version of the capacitated fault-tolerant  $\kappa$ -center in which every vertex has capacity either zero or  $L$ . Vertices with capacity 0 are called *0-vertices* and vertices with capacity  $L$  are called *L-vertices*. For a given set  $A$  of vertices, we denote by  $A^L$  the set containing all  $L$ -vertices of  $A$ .

### 6.1 LP-Formulation

We give a rounding algorithm for the  $\{0, L\}$ -capacitated case. As in Sect. 5, we formulate the problem using  $ILP_{k,\alpha}(G)$ . In this case, however, we may rewrite the program such that only  $L$ -vertices appear in the summation, and all coefficients are equal, that is,  $ILP_{k,\alpha}(G)$  can be written as:

$$\begin{aligned} \sum_{u \in V} y_u &= k \\ |U| &\leq \sum_{u \in (N_G(U))^L \setminus F} y_u L & \forall U \subseteq V, F \subseteq V : |F| = \alpha \\ y_u &\in \{0, 1\} & \forall u \in V. \end{aligned}$$

Notice that the second line in the program above can be simplified. The key observation is that, in the worst case, the total failed capacity is always the constant  $\alpha L$ . Indeed, consider a feasible integer solution  $y$  and a fixed subset  $U \subseteq V$  such that  $U \neq \emptyset$ , and let  $H = \{u \in (N_G(U))^L : y_u = 1\}$ . We have  $|H| > \alpha$ , since otherwise

we would get  $|U| \leq \sum_{u \in (N_G(U))^L \setminus H} y_u L = \sum_{u \in (N_G(U))^L \setminus H} 0 \cdot L = 0$ , that is a contradiction since  $U$  is not empty. Let  $F'$  be any subset of  $H$  with  $|F'| = \alpha$ . From the inequality constraint in  $ILP_{k,\alpha}(G)$  for  $F = F'$ , we obtain

$$\begin{aligned} |U| &\leq \sum_{u \in (N_G(U))^L \setminus F'} y_u L = \sum_{u \in (N_G(U))^L} y_u L - \sum_{u \in (N_G(U))^L \cap F'} 1 \cdot L \\ &= \sum_{u \in (N_G(U))^L} y_u L - \alpha L. \end{aligned}$$

Therefore, the following linear program, that is denoted by  $LPU_{k,\alpha}(G)$ , is a relaxation of  $ILP_{k,\alpha}(G)$ .

$$\begin{aligned} \sum_{u \in V} y_u &= k \\ |U| &\leq \sum_{u \in (N_G(U))^L} y_u L - \alpha L & \forall U \subseteq V, U \neq \emptyset \\ 0 &\leq y_u \leq 1 & \forall u \in V. \end{aligned}$$

In contrast to  $LP_{k,\alpha}(G, \{C_v\}_{v \in V})$ , this program can be separated even if  $\alpha$  is part of the instance. The difference is that, in this formulation, the failure scenarios need not be enumerated. Given a candidate solution  $y$ , we can compute the minimum value of  $\sum_{u \in N^L(U)} y_u L - |U|$  over all sets  $U$ , and check whether this value is at least  $\alpha L$ . This can be done in polynomial time using a max-flow min-cut algorithm with arguments very similar to those in the proof of Lemma 6. This means that we can separate  $LPU_{k,\alpha}(G)$  in polynomial time, which implies the following lemma.

**Lemma 9**  *$LPU_{k,\alpha}(G)$  can be solved in polynomial time even if  $\alpha$  is part of the input.*

## 6.2 Rounding

For the non-fault-tolerant  $\{0, L\}$ -capacitated  $\kappa$ -center, An et al. [10] perform an additional preprocessing of the input graph to obtain a clustering with stronger properties. This way, they derive an integral distance-5 transfer. Namely, before the preprocessing described in Sect. 2, which produces an unweighted connected graph  $G = (V, E)$ , they remove any edge connecting two 0-vertices. We apply their rounding algorithm to a solution of  $LPU_{k,\alpha}(G)$ , which leads to the following result.

**Lemma 10** *Suppose  $G = (V, E)$  is a connected graph such that each vertex is either a 0-vertex or an  $L$ -vertex, no two adjacent vertices are 0-vertices, and  $y : V \rightarrow [0, 1]$  is a vector such that  $\sum_{u \in (N(v))^L} y_u \geq 1$  for  $v \in V$  and  $\sum_{v \in V} y_v$  is an integer. Then there is a polynomial-time algorithm that produces an integral distance-5 transfer  $y'$  of  $y$ .*

Now we obtain a 6-approximation for the  $\{0, L\}$ -capacitated case.

**Theorem 4** *There exists a 6-approximation for the  $\{0, L\}$ -capacitated  $\alpha$ -fault-tolerant  $\kappa$ -center (with  $\alpha$  as part of the input).*

*Proof* Let  $y$  be an optimal solution for  $LPU_{k,\alpha}(G)$ , and  $y'$  be an integral distance-5 transfer of  $y$  obtained by the algorithm of Lemma 10. Also, let  $R$  be the set of centers



corresponding to the characteristic vector  $y'$ . We proceed as in the proof of Lemma 8. Consider a subset  $U \subseteq V$ . Let  $X(U) = \{v : y_v > 0 \text{ and } v \in (N_G(U))^L\}$ , and let  $Y(U) = N_G^5(X(U)) \cap R$  be the set of all vertices whose  $y'$  values might have been partially transferred, by the algorithm of Lemma 10, from the  $y$  values of vertices in  $X(U)$ . By the constraints of  $LPU_{k,\alpha}(G)$  and the fact that  $y'$  is an integral transfer, we get

$$|U| + \alpha L \leq \sum_{u \in X(U)} y_u L \leq \sum_{u \in Y(U)} y'_u L = |Y(U)|L.$$

Now consider a failure scenario  $F \subseteq V$  with  $|F| = \alpha$ . We can create a bipartite graph (as in Lemma 8) that connects each vertex  $u \in V$  to vertices  $Y(\{u\}) \setminus F \subseteq R$ . Using Hall's condition, we obtain an assignment  $\phi : V \rightarrow R \setminus F$  that respects the capacities. Since  $y'$  is a distance-5 transfer, we know that  $Y(\{u\}) \subseteq N^6(\{u\})$  for every  $u$ , and thus  $d(u, \phi(u)) \leq 6$ .  $\square$

## 7 The $k$ -Supplier

In this section, we consider the  $\kappa$ -supplier problem. In this problem, one is given a set  $\mathcal{C}$  of clients, a set  $\mathcal{F}$  of candidate locations for facilities, and an integer  $k$ . The goal is to select  $k$  of the locations to install facilities to serve each of the clients so that the maximum distance between a client and its assigned facility is minimized.

It is easy to see that  $\kappa$ -center is the special case of  $\kappa$ -supplier in which  $\mathcal{C} = \mathcal{F}$ . So, in particular, any approximation for  $\kappa$ -supplier applies also to  $\kappa$ -center and achieves the same approximation factor. The same holds for the capacitated and for the fault-tolerant versions of the problems. For the basic  $\kappa$ -supplier, there is a 3-approximation, whose factor is the best possible unless  $P = NP$  [5]. Also, there exists a 3-approximation [14] for the fault-tolerant  $\kappa$ -supplier, and an 11-approximation for the capacitated  $\kappa$ -supplier [10].

In the capacitated fault-tolerant version of the  $\kappa$ -supplier, each client must be assigned to a facility, even at the failure of up to  $\alpha$  facilities, and the assignment is such that no facility  $u$  is assigned more than  $L_u$  clients. In the following, we show that our algorithms for the  $\kappa$ -center problem naturally extend to this generalization, for both the conservative and the non-conservative variants.

Table 2 summarizes the obtained approximation factors.

As in the case of the  $\kappa$ -center problem, we reduce the problem to the case of a unweighted connected graph  $G$  and the objective is to obtain a distance-1 solution.

**Table 2** Summary of the obtained approximation factors for the  $\kappa$ -supplier problem

Version	Capacities	Value of $\alpha$	Factor
Conservative	Uniform	Given in the input	7
Conservative	Arbitrary	Fixed	$11 + 8\alpha$
Non-conservative	Uniform	Given in the input	7
Non-conservative	Arbitrary	Fixed	13

For the  $\kappa$ -supplier problem, however, we consider only edges between  $\mathcal{C}$  and  $\mathcal{F}$ , that is, the obtained graph is bipartite. This implies that distances in  $G$  between pairs of clients or between pairs of facilities are even.

## 7.1 The Non-conservative Capacitated Fault-Tolerant $k$ -Supplier

We first consider the case that capacities are non-uniform. A slightly different formulation from  $ILP_{k,\alpha}(G)$  is used: the main difference is that we only have variables  $y_u$  for elements  $u$  of  $\mathcal{F}$ , and we only consider constraints corresponding to subsets of clients  $U \subseteq \mathcal{C}$  and failure scenarios  $F \subseteq \mathcal{F}$ .

By adapting the example of Sect. 5.1, the obtained formulation also has unbounded integrality gap, and thus we consider a relaxation based on a modified graph that depends on a clustering. In this step, rather than selecting clients at distance 3, we greedily pick clients whose distance to previously picked elements is exactly 4. This set of elements  $\Gamma$  (midpoints) induces a clustering of  $\mathcal{F}$ , and a corresponding tree of midpoints such that any adjacent midpoints in the tree are at distance 4, and every facility is associated to a midpoint at distance at most 3.

As in the case of  $\kappa$ -center, we select a set  $B_v$  of  $\alpha$  facilities of largest capacity in each cluster centered at  $v \in \Gamma$ , and construct a graph  $G'$  by adding arcs from any client at distance at most 2 from a midpoint  $v$  to each facility of  $B_v$ . Let  $B$  be the union of all  $B_v$  for  $v \in \Gamma$ . The obtained LP relaxation is:

$$\begin{aligned} \sum_{u \in \mathcal{F}} y_u &= k \\ |U| &\leq \sum_{u \in N_{G'}(U) \setminus F} y_u L_u \quad \forall U \subseteq \mathcal{C}, F \subseteq B : |F| = \alpha \\ 1 &\leq \sum_{u \in N_G(v) \setminus B} y_u \quad \forall v \in \Gamma \\ y_u &= 1 \quad \forall u \in B \\ 0 &\leq y_u \leq 1 \quad \forall u \in V. \end{aligned}$$

As done in [10], a rounding algorithm similar to that for  $\kappa$ -center can obtain an integral distance-10 transfer of a solution for the previous linear program (the only difference is that a distance-2 transfer on the tree of midpoints is now interpreted as a distance- $(2 \cdot 4)$  transfer on the original graph). This transfer implies that, for a failure scenario  $F \subseteq B$ , one may obtain an assignment  $\phi$  such that  $d(u, \phi(u)) \leq 11$  for every  $u \in \mathcal{C}$ . Moreover, by using the same reasoning as in the proof of Lemma 8, one may show that  $d(u, \delta(\phi(u))) \leq 10$ , where  $\delta(\phi(u))$  is the midpoint associated with  $\phi(u)$ . Therefore, the distance-11 assignment for a failure scenario  $F' \subseteq B$  can be transformed into a distance-13 assignment for a general failure scenario  $F \subseteq \mathcal{F}$ .

For the uniformly capacitated case, we can also obtain a simplified relaxation as in Sect. 6. It is possible to adapt the rounding algorithm for the  $\{0, L\}$ -capacitated  $\kappa$ -center by An et al. [10], and obtain an integral distance-6 transfer for the solution for this relaxation. The key observation is that, since the underlying graph is bipartite, we can interpret clients as 0-vertices, so that no two such vertices are adjacent. This leads to the following lemma.

**Lemma 11** Suppose  $G = (\mathcal{C} \cup \mathcal{F}, E)$  is a connected bipartite graph such that each  $v \in \mathcal{F}$  has capacity  $L$ , and  $y : V \rightarrow [0, 1]$  is a vector such that  $\sum_{u \in N(v)} y_u \geq 1$  for  $v \in \mathcal{C}$  and  $\sum_{v \in V} y_v$  is an integer. Then there is a polynomial-time algorithm that produces an integral distance-6 transfer  $y'$  of  $y$ .

The reason that the algorithm obtains a distance-6 transfer for the  $\kappa$ -supplier, rather than a distance-5 transfer, is that cluster midpoints are at distance 4 in an instance of the  $\kappa$ -supplier, whilst midpoints are at distance 3 in an instance of  $\kappa$ -center. Now, using Lemma 11 and repeating the arguments in the proof of Theorem 4, we obtain a 7-approximation for the uniformly capacitated fault-tolerant  $\kappa$ -supplier.

## 7.2 The Conservative Capacitated Fault-Tolerant $k$ -Supplier

First, we revisit the notion of independent sets for the  $\kappa$ -supplier. A set of *facilities*  $W$  is  $(\alpha, \ell)$ -independent if each connected component of  $G^{\ell-1}[W]$  contains at most  $\alpha$  vertices. Also, a set of *clients*  $A$  is 8-independent if  $d(u, v) \geq 8$  for every  $u, v \in A$  (notice that, in this bipartite setting, requiring that a set of clients is 7-independent is the same as requiring that it is 8-independent). With these adapted definitions, one may obtain versions of Lemmas 1 and 2 with analogous statements.

For the uniformly capacitated case, we use Algorithm 1, but with an 8-independent set  $A$ , and assuming that ALG is a  $\beta$ -approximation for the capacitated  $\kappa$ -supplier problem. Notice that since  $A$  is maximal, for every client  $u$ , there is a client  $v \in A$ , such that  $d(u, v) \leq 6$ . Now, by repeating Theorem 1, we obtain that this algorithm has approximation factor  $\max\{7, \beta\}$ . We use the algorithm by An et al. for the uniformly capacitated case (without failures), for which, as stated above,  $\beta = 7$ , and obtain a 7-approximation.

For the non-uniformly capacitated case, we use Algorithm 2. However, when augmenting the set of backup facilities  $B$  with a set of facilities  $U$  (Line 6), rather than excluding elements in  $N^6(U) \cap B$ , we exclude the elements in  $N^8(U) \cap B$ . Recall that, in the  $\kappa$ -center problem, we obtain a 7-independent set  $A \subseteq B$  by selecting an element in each connected component of  $G^6[B]$  (see the proof of Lemma 3). In the case of the  $\kappa$ -supplier problem, to obtain an 8-independent set  $A$  of clients, we must choose from the neighborhood of the set  $B$  of backup facilities (and not directly from  $B$ ). Thus, for each connected component  $C_i$  of  $G^8[B]$ , we choose a facility  $b_i \in C_i$ , and a neighbor  $a_i \in N(b_i) \subseteq \mathcal{C}$ . Notice that, for any pair  $b_i, b_j$ ,  $d(b_i, b_j) > 8$ , thus  $d(b_i, b_j) \geq 10$ , and hence  $d(a_i, a_j) \geq 8$ . Therefore, the set  $A$  of all  $a_i$ 's is an 8-independent set. The rest of the proof remains unchanged, except that we replace 6 by 8, obtaining a factor  $\beta + 8\alpha$ . The best known approximation for the capacitated  $\kappa$ -supplier has factor  $\beta = 11$  [10].

## 8 Complexity Results

In Algorithm 2, one wants to decide whether or not, for every set of vertices  $U$  with up to  $\alpha$  vertices, it is the case that  $L(U) \leq L(B \cap N^6(U))$ . The next theorem shows that this problem is coNP-complete when  $\alpha$  is part of the input. First, we notice that

we may consider  $N(U)$  instead of  $N^6(U)$ , as there is a reduction from the problem defined using  $N(U)$  to the problem defined using  $N^6(U)$ .

**Lemma 12** *Let  $G$  be a graph,  $B \subseteq V(G)$ ,  $\alpha \geq 0$ , and  $L_u \geq 0$  for  $u \in V$ . Also, let  $H$  be a graph obtained from  $G$  by replacing each edge  $\{u, v\}$  by a path of length 6 connecting  $u$  to  $v$ , and defining  $L_u = 0$  for each added vertex. There exists  $U \subseteq V(G)$  with  $|U| \leq \alpha$  and  $L(U) > L(B \cap N_G(U))$  if and only if there exists  $U \subseteq V(H)$  with  $|U| \leq \alpha$  and  $L(U) > L(B \cap N_H^6(U))$ .*

*Proof* First suppose there exists  $U \subseteq V(G)$  with  $|U| \leq \alpha$  and  $L(U) > L(B \cap N_G(U))$ . By construction of  $H$ , we have  $L(B \cap N_G(U)) = L(B \cap N_H^6(U))$ . It follows that  $L(U) > L(B \cap N_H^6(U))$ .

For the opposite direction, suppose there exists  $U \subseteq V(H)$  with  $|U| \leq \alpha$  and  $L(U) > L(B \cap N_H^6(U))$ . We obtain  $L(U \cap V(G)) = L(U) > L(B \cap N_H^6(U)) \geq L(B \cap N_H^6(U \cap V(G))) = L(B \cap N_G(U \cap V(G)))$ .  $\square$

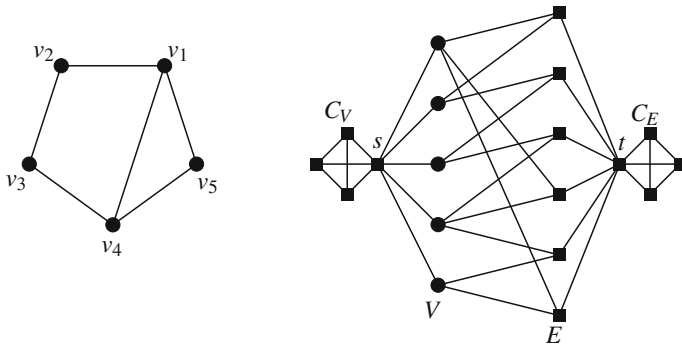
**Theorem 5** *The problem of, given a graph  $H = (V_H, E_H)$ , a number  $L_u$  for each  $u \in V_H$ , a set  $B \subseteq V_H$ , and a number  $\alpha$ , deciding whether  $L(U) \leq L(B \cap N_H(U))$  for every  $U \subseteq V_H$  with  $|U| \leq \alpha$  is coNP-complete. Moreover, this problem, when parameterized by  $\alpha$ , is  $W[1]$ -hard.*

*Proof* This problem is in coNP, because, for an instance  $(H, L, B, \alpha)$  whose answer to the problem is no, that is, a NO instance, one can present as a NO certificate a set  $U \subseteq V_H$  such that  $|U| \leq \alpha$  and  $L(U) > L(B \cap N_H(U))$ .

The clique problem is known to be NP-complete [18] and consists in, given a graph  $G$  and a positive integer  $k$ , to decide whether there exists a clique in  $G$  with at least  $k$  vertices. We present a reduction from the clique problem to our problem so that an instance  $(G, k)$  of the clique problem is a YES instance if and only if the corresponding instance  $(H, L, B, \alpha)$  for our problem is a NO instance.

Let  $(G, k)$  be an instance of the clique problem with  $G = (V, E)$ . The main part of the graph  $H$  consists of the bipartite graph with bipartition  $\{V, E\}$  and a vertex  $v$  in  $V$  adjacent to an edge  $e$  in  $E$  if  $v$  is an end of  $e$  in  $G$ . Besides this, graph  $H$  has two disjoint cliques on  $k + 1$  vertices, say  $C_V$  and  $C_E$ . A vertex in  $C_V$ , say  $s$ , is adjacent to each vertex in  $V$  while a vertex in  $C_E$ , say  $t$ , is adjacent to each edge in  $E$ . This finishes the description of graph  $H$ . See Fig. 5 for an example. The capacity function  $L$  is defined as follows. For each  $e$  in  $E$ , let  $L(e) = 1$ ; for each  $v$  in  $V$ , let  $L(v)$  be the degree of  $v$  in  $G$ , denoted as  $d_v$ ; for each  $u$  in  $C_V$ , let  $L(u) = \binom{k}{2} - 1$  and, for each  $u$  in  $C_E$ , let  $L(u) = |E|$ . Finally, let  $B = E \cup C_V \cup C_E$  and  $\alpha = k$ . This concludes the description of the instance of our problem, which can be obtained from  $(G, k)$  in time polynomial in the size of  $(G, k)$ . Next we show that  $(G, k)$  is a YES instance of the clique problem if and only if  $(H, L, B, \alpha)$  is a NO instance of our problem.

First let us prove that, if there exists a clique  $S$  of size  $k$  in  $G$ , then  $L(S) > L(B \cap N_H(S))$ , that is, the answer of our problem for the instance  $(H, L, B, \alpha)$  is NO. Indeed,  $B \cap N_H(S)$  consists of the special vertex  $s$  in  $C_V$  and the edges incident to  $S$  in  $G$ , so  $L(B \cap N_H(S)) = \binom{k}{2} - 1 + \ell$ , where  $\ell$  is the number of edges in  $G$  incident to  $S$ . The value of  $L(S)$  is  $\sum_{v \in S} d_v$ , which is exactly the number of edges incident to  $S$  plus the number of edges in  $G$  with both ends in  $S$ , that is, the edges in



**Fig. 5** The graph on the right is the graph  $H$  in the instance of our problem corresponding to the graph  $G$  on the left for  $k = 3$ . Squared vertices indicate the set  $B$

the graph  $G[S]$  induced by  $S$ . As the number of edges in  $G[S]$  is exactly  $\binom{k}{2}$  because  $S$  is a clique on  $k$  vertices,  $L(S) = \binom{k}{2} + \ell > L(B \cap N_H(S))$ , as we wished.

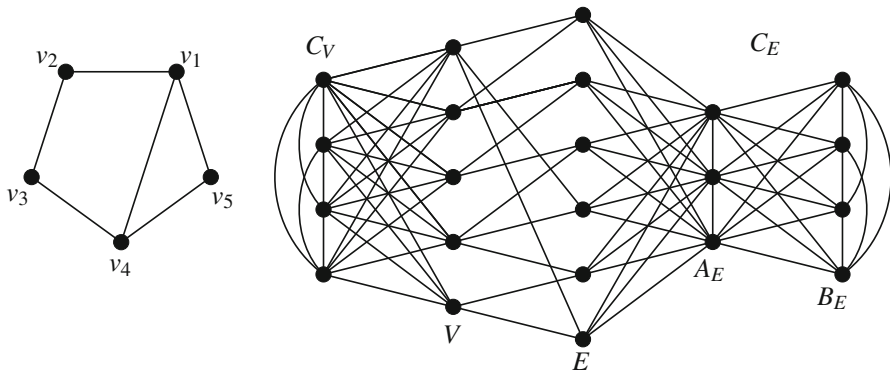
Second we prove that, if  $L(U) > L(B \cap N_H(U))$  for a set  $U$  of up to  $k$  vertices of  $H$ , then there is a clique with  $k$  vertices in  $G$ . We start by arguing that  $L(U \cap (C_V \cup C_E)) \leq L((B \cap N_H(U) \cap (C_V \cup C_E)) \setminus \{s, t\})$ . If  $U \cap C_V \neq \emptyset$ , then  $B \cap N_H(U) \supseteq C_V$ . Moreover,  $U \neq C_V$  since  $C_V$  has  $k + 1$  vertices and  $U$  has up to  $k$  vertices. This means that  $L(U \cap C_V) \leq L((B \cap N_H(U) \cap C_V) \setminus \{s\})$ . Similarly, if  $U \cap C_E \neq \emptyset$ , then  $B \cap N_H(U) \supseteq C_E$ . Again  $U \neq C_E$ , so we have that  $L(U \cap C_E) \leq L((B \cap N_H(U) \cap C_E) \setminus \{t\})$ , completing the proof of the claimed inequality. Now note that  $L(U \cap E) \leq |E| = L(t)$ . On the other hand, let  $S = U \cap V$  and note that  $L(S) = \sum_{v \in S} d_v$ , which is exactly the number of edges incident to  $S$  plus the number of edges in the graph  $G[S]$ . If  $S$  is not a clique on  $k$  vertices, then  $L(S) \leq L(E \cap N_H(U)) + L(s)$  and, joining everything, we deduce that  $L(U) \leq L(B \cap N_H(U))$ , a contradiction. So  $S$  must be a clique on  $k$  vertices in  $G$ .

To check  $W[1]$ -hardness, it suffices to observe that the reduction takes polynomial time in the size of  $G$ ,  $k \leq |V|$ , and  $\alpha = k$ . Since clique is  $W[1]$ -hard [19], the result follows.  $\square$

Analogously, we reduce the separation problem for program  $LP_{k,\alpha}(G, \{C_v\}_{v \in \Gamma})$  to a related problem. In the following theorem, we show that this problem is coNP-hard when  $\alpha$  is part of the input. Thus, to achieve a constant approximation for capacitated fault-tolerant  $\kappa$ -center with general capacities and  $\alpha$  as part of input, one is likely to need a different strategy.

**Theorem 6** *The problem of, given a graph  $H = (V_H, E_H)$ , a number  $L_u$  for each  $u \in V_H$ , and a number  $\alpha$ , deciding whether  $L(N_H(U) \setminus F) \geq |U|$  for every  $U \subseteq V_H$  and  $F \subseteq V_H$  with  $|F| = \alpha$  is coNP-complete. Moreover, this problem, when parameterized by  $\alpha$ , is  $W[1]$ -hard.*

*Proof* The proof is similar to that of Theorem 5. It is easy to see that the problem is in coNP as, for a NO instance of the problem, one can present as a certificate subsets  $U$  and  $F$  of  $V_H$  such that  $|F| = \alpha$  and  $L(N_H(U) \setminus F) < |U|$ .



**Fig. 6** The graph on the right is the graph  $H$  in the instance of our problem corresponding to the graph on the left for  $k = 3$

Consider again the NP-complete clique problem: given a graph  $G$  and a positive integer  $k$ , decide whether there exists a clique in  $G$  with at least  $k$  vertices. Next we present a reduction from the clique problem to our problem so that an instance  $(G, k)$  of the clique problem is a YES instance if and only if the corresponding instance  $(H, L, \alpha)$  for our problem is a NO instance.

Let  $(G, k)$  be an instance of the clique problem, where  $G = (V, E)$ . The main part of the graph  $H$  consists of the bipartite graph with  $V$  as one side and  $E$  as the other side of the bipartition. A vertex  $v$  in  $V$  is adjacent to an edge  $e$  in  $E$  if  $v$  is an end of  $e$  in  $G$ . Besides this,  $H$  has two disjoint cliques, say,  $C_V$  on  $k + 1$  vertices and  $C_E = A_E \cup B_E$  on  $2k + 1$  vertices, with  $|A_E| = k$  and  $|B_E| = k + 1$ . Every vertex in  $C_V$  is adjacent to each vertex in  $V$  and every vertex in  $A_E$  is adjacent to each edge in  $E$ . This finishes the description of graph  $H$ . See Fig. 6 for an example. As for  $L$ , for each  $e$  in  $E$ , let  $L(e) = 0$ ; for each  $v$  in  $V$ , let  $L(v) = d_v$ , where  $d_v$  is the degree of  $v$  in  $G$ ; for each  $u$  in  $C_V \cup B_E$ , let  $L(u) = |V_H|$ ; denoting by  $a_1, \dots, a_k$  the vertices in  $A_E$ , let  $L(a_i) = i$  for  $i = 1, \dots, k - 1$  and  $L(a_k) = k - 1$ . Finally, let  $\alpha = k$ , concluding the description of the instance of our problem, which can be obtained from  $(G, k)$  in time polynomial in the size of  $(G, k)$ . Next we show that  $(G, k)$  is a YES instance of the clique problem if and only if  $(H, L, \alpha)$  is a NO instance of our problem.

First, suppose that there exists a clique  $S$  of size  $k$  in  $G$ . Let  $U$  be the edges in  $G[S]$  and let  $F = S$ . Note that  $|F| = |S| = k = \alpha$  and  $|U| = \binom{k}{2}$ , as  $S$  is a clique on  $k$  vertices. Thus  $N_H(U) \setminus F = (S \cup A_E) \setminus F = A_E$ , and  $L(N_H(U) \setminus F) = L(A_E) = \binom{k}{2} - 1 < |U|$ . Hence the answer of our problem for the instance  $(H, L, \alpha)$  is NO.

Second, suppose that there are subsets  $U$  and  $F$  of  $V_H$  such that  $|F| = \alpha$  and  $L(N_H(U) \setminus F) < |U|$ . Observe that  $U \cap (V \cup C_V) = \emptyset$ ; otherwise,  $N_H(U) \supseteq C_V$  and  $C_V \setminus F \neq \emptyset$  because  $F$  has  $k$  vertices and  $C_V$  has  $k + 1$  vertices. But this would mean that  $L(N_H(U) \setminus F) \geq |V_H| \geq |U|$ , a contradiction. Similarly  $U \cap C_E = \emptyset$ ; otherwise,  $N_H(U) \supseteq B_E$  and, as  $B_E$  has  $k + 1$  vertices,  $L(N_H(U) \setminus F) \geq |V_H| \geq |U|$ , a contradiction. So we know that  $U \subseteq E$ . Now let  $S = N_H(U) \cap V$  and  $\ell = |S \cap F|$ . Thus  $N_H(U)$  has at least  $\ell$  vertices in  $A_E \setminus F$ , and then  $L(N_H(U) \setminus F) \geq L(\{a_1, \dots, a_\ell\}) + L(S \setminus F)$ . Notice that  $L(\{a_1, \dots, a_\ell\}) = \binom{\ell}{2}$  if  $\ell < k$ , and  $L(\{a_1, \dots, a_\ell\}) = \binom{\ell}{2} - 1$  if  $\ell = k$ .

Let  $U'$  be the set of all edges that have both ends in  $S \cap F$ . It follows that every edge in  $U \setminus U'$  has at least one end in  $S \setminus F$ . On the one hand,  $L(S \setminus F) = \sum_{u \in S \setminus F} d_u$ , which is the number of edges incident to  $S \setminus F$  plus the number of edges in  $G[S \setminus F]$ , thus  $L(S \setminus F) \geq |U \setminus U'|$ . On the other hand, every edge of  $U'$  is in  $G[S \cap F]$  so, since  $\ell = |S \cap F|$ , the graph  $G[S \cap F]$  contains at most  $\binom{\ell}{2}$  edges, and thus  $|U'| \leq \binom{\ell}{2}$ . We obtain  $|U| > L(N_H(U) \setminus F) \geq L(\{a_1, \dots, a_\ell\}) + L(S \setminus F) \geq L(\{a_1, \dots, a_\ell\}) + |U \setminus U'|$ . But then  $\binom{\ell}{2} - 1 \leq L(\{a_1, \dots, a_\ell\}) \leq |U| - |U \setminus U'| - 1 \leq |U'| - 1 \leq \binom{\ell}{2} - 1$ . Since  $L(\{a_1, \dots, a_\ell\}) = \binom{\ell}{2} - 1$ , we deduce that  $\ell = k$ . We conclude that  $|U'| = \binom{k}{2}$ , and thus  $G[S \cap F]$  is a clique on  $k$  vertices.

To check  $W[1]$ -hardness, observe that the reduction takes polynomial time in the size of  $G$ ,  $k \leq |V|$ , and  $\alpha = k$ .  $\square$

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