

# A polynomial time approximation scheme for embedding a directed hypergraph on a ring

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## Abstract

We study the problem of embedding a directed hypergraph on a ring that has applications in optical network communications. The undirected version (MCHEC) has been extensively studied. It was shown that the undirect version was NP-complete. A polynomial time approximation scheme (PTAS) for the undirected version has been developed. In this paper, we design a polynomial time approximation scheme for the directed version.

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## 1. Introduction

The problem of embedding hyperedges on a ring was originally proposed for electronic design automation, where the objective is to route within a minimum-area rectangle [9,1]. The problem of embedding undirected hyperedges on a ring with minimum congestion (MCHEC) has applications in parallel computing as well as multicast routing. Here we study the problem of embedding directed hyperedges on a ring. It models the case where the links in the network is directed.

For the undirected case (MCHEC), Ganley and Co-hoon in [2] proved that the problem was NP-hard and

gave a ratio-3 approximation algorithm. They also gave an algorithm that can solve the case where the congestion is a constant. Several ratio-2 approximation algorithms were given in [3,5]. Gu and Wang presented a ratio-1.8 approximation algorithm [4]. Recently, Deng and Li proposed a polynomial time approximation scheme (PTAS) for the problem. Their approach comes from the techniques for string problems [6].

In this paper, we study the problem of embedding directed hyperedges on a ring. We extend the method in [8] to get a polynomial time approximation scheme for the directed version. We have developed a new technique for the proofs of some key lemmas. This technique can also be applied to the undirected case. The new proofs allow us to reduce the time complexity of the algorithms in [8] by a factor of  $O(m)$ , where  $m$  is the total number of hyperedges.

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The paper is organized as follows: We first give some definitions in Section 2. Section 3 deals with a special case, where the number of hyperedges  $m$  is  $O(\log n)$ . Section 4 gives an algorithm that solves the case where the number of hyperedges  $m$  is of  $O(C_{\text{opt}})$ , and  $c_{\text{opt}}$  is the minimum congestion cost of an optimal embedding. The general case is solved in Section 5.

## 2. Preliminaries

A *ring* of  $n$  nodes is a directed graph  $R = (V, E_R)$ , where  $V = \{1, 2, \dots, n\}$  is the set of  $n$  vertices on the ring and  $E_R = \{e_i^+ = (i, i+1), e_i^- = (i+1, i) \mid i = 1, 2, \dots, n\}$  is the set of  $2n$  directed edges on the ring, where  $n+1$  is treated as 1. Consider the same set of vertices  $V = \{1, 2, \dots, n\}$ . A directed hyperedge  $h = (u, S)$  is a pair, where  $u \in V$  is the source of the hyperedge and  $S \subseteq V - \{u\}$  is the set of sinks. In communication applications, each hyperedge  $h$  represents a request that asks to send a message from  $u$  to every vertex in  $S$ . Let  $H = (V, E_H)$  be a directed hypergraph with the same set of vertices  $V$  and a set of  $m$  directed hyperedges  $E_H = \{h_1, h_2, \dots, h_m\}$ .

Let  $h_j = (u_j, S_j)$  be a hyperedge in  $E_H$ , where  $S_j = \{i_1^j, i_2^j, \dots, i_{k_j}^j\}$ .  $k_j = |S_j|$  denotes the total number of sink vertices in the hyperedge. For convenience, we use  $i_0^j$  to denote  $u_j$ . Assume that the  $k_j + 1$  vertices  $i_0^j, i_1^j, \dots, i_{k_j}^j$  follow the clockwise order on the ring.  $P_k^j$  denotes the segment of vertices on the ring from vertex  $i_k^j$  to vertex  $i_{k+1}^j$  for  $k = 0, 1, \dots, k_j - 1$  and  $P_{k_j}^j$  denotes the segment of vertices on the ring from vertex  $i_{k_j}^j$  to vertex  $i_0^j$ . In order to realize the request  $h_j$  on the ring, one can cut one of the paths  $P_k$  for  $k = 0, 1, \dots, k_j$  and obtain two directed paths on the ring both starting from  $i_0^j$ . This forms an embedding of  $h_j$  on the ring. For a hyperedge  $h_j$ , there are  $k_j + 1$  different embeddings, one for each  $P_k^j$  (by cutting  $P_k^j$ ). An *embedding* of  $h_j$  is a  $P_k^j$  *embedding* if  $P_k^j$  is cut.

Given an embedding  $x$  of all the hyperedges  $E_H$ , the *congestion*  $e_i^+(x)$  or  $e_i^-(x)$  of a directed edge  $e_i^+$  or  $e_i^-$  is the number of times that the edge  $e_i^+$  or  $e_i^-$  is used in the embedding. When  $x$  is clear, we also use  $c(e_i^+)$  and  $c(e_i^-)$  to denote the congestion. The problem here is to find an embedding for each  $h_j \in E_H$  such that every edge  $e_i^+$  and  $e_i^-$  on the ring is used at most  $c$  times and  $c$  is minimized. We refer the problem as the *embedding directed hyperedges on the ring* problem (EDHR, for short).

## 3. Enumerating method for $m = O(\log n)$

In this section, we give an algorithm with ratio  $1 + \frac{1}{r}$  for the case where there are  $O(\log n)$  hyperedges in  $E_H$ . The basic idea of our algorithm is to choose  $2r$  edges on the ring, where  $r$  is a constant related to the ratio, and for each hyperedge  $h_j$  we only have to cut a  $P_k^j$  that contains one of the  $2r$  selected edges on the ring. By doing this, for each  $h_j$  we only have to consider  $2r$  choices (instead of  $k_j + 1$  choices for an optimal solution). Since there are at most  $O(\log n)$  hyperedges, the time complexity is  $O((2r)^{O(\log n)})$ , that is polynomial.

Let  $i_1, i_2, \dots, i_{2r}$  be  $2r$  indices representing the  $2r$  edges  $(i_1, i_1 + 1), (i_2, i_2 + 1), \dots, (i_{2r}, i_{2r} + 1)$  on the ring  $R$ . Let  $x = (x_1, x_2, \dots, x_m)$  be an embedding of  $H$ , where  $x_j$  indicates the choice  $P_k^j$  that is cut for the embedding of  $h_j$ . We use  $E_j(x)$  to denote the segment  $P_k^j$  that is cut for the embedding  $x$  of  $h_j$ .

Let  $Q_{i_1, i_2, \dots, i_{2r}}(x)$  be a set of indices of hyperedges such that  $j$  is in  $Q_{i_1, i_2, \dots, i_{2r}}(x)$  if and only if  $E_j(x)$  contains at least one of the  $2r$  edges  $(i_1, i_1 + 1), (i_2, i_2 + 1), \dots, (i_{2r}, i_{2r} + 1)$ .

**Lemma 1.** *Let  $x$  be any embedding of  $H$ . For any fixed index  $1 \leq i_1 \leq n$ , there exist  $2r - 1$  indices  $i_2, i_3, \dots, i_{2r}$  such that for any embedding  $x'$  satisfying  $x'_j = x_j$  for every  $j \in Q_{i_1, i_2, \dots, i_{2r}}(x)$ , we have*

$$e_i^+(x') - e_i^+(x) \leq \frac{1}{r} e_i^+(x) \quad \text{and}$$

$$e_i^-(x') - e_i^-(x) \leq \frac{1}{r} e_i^-(x)$$

for any directed edge  $e_i^+$  and  $e_i^-$  in  $E_R$  on the ring.

**Proof.** We prove the lemma by giving a way to find the  $2r - 1$  indices. Let  $c$  be the congestion for the embedding  $x$  that we want to approximate. First, we select an arbitrary edge, say,  $e_1^+ = (1, 2) = (i_1, i_1 + 1)$  on the ring. In the embedding of  $x$ , there are at least  $m - 2c$   $h_j$ 's with  $e_1^+ \in E_j(x)$ . That is, there are at most  $2c$   $h_j$ 's with  $e_1^+ \notin E_j(x)$ . Let  $H_r$  be the set of the (at most  $2c$ )  $h_j$ 's with  $e_1^+ \notin E_j(x)$ . We use the following method to select the remaining (at most)  $2r - 1$  indices.

1. **for** a remaining edge  $e_g^+$  on the ring  $R$  **do**
2. **if** there are more than  $\frac{c}{r}$  hyperedges  $h_j \in H_r$  with  $e_g^+ \in E_j(x)$ , **then** we select the index  $q$  and set  $H_r = H_r - \{j \mid e_g^+ \in E_j(x)\}$ .
3. **if** the size of  $H_r$  is more than  $\frac{c}{r}$  **then** goto Step 1 **else** stop.

The above procedure will stop after at most  $2r - 1$  iterations since each time the size of  $H_r$  is reduced by at least  $\frac{c}{r}$  and the original size of  $H_r$  is at most  $2c$ .

Now, consider any edge  $e_i^+$  or  $e_i^-$  with index  $i$  ( $1 \leq i \leq n$ ) not selected in the above procedure. For the embedding  $x$ , the number of  $h_j$ 's that are not cut at edge  $e_i^+$  or  $e_i^-$  in  $x$  is at most the size of  $H_r$ , that is upper bounded by  $\frac{c}{r}$ . Thus, the lemma holds.  $\square$

Note that, in the proof of Lemma 1, we assume that  $x$  is known when selecting the  $2r - 1$  indices. In fact,  $x$  will be the optimal solution that we do not know. However, we can go through all possible sets of  $2r - 1$  indices in polynomial time. Based on Lemma 1, we can solve the problem as follows:

1. try all possible choices of  $i_2, i_3, \dots, i_{2r}$ .
2. for each  $h_j \in H$ , try the  $2r - 1$  choices for cutting the path  $P_k^j$  containing  $e_{i_1}^+, e_{i_2}^+, \dots$ , or  $e_{i_{2r}}^+$ .

Step 1 takes  $O(n^{2r-1})$  time and Step 2 needs  $O((2r)^m) = O((2r)^{O(\log n)}) = n^{O(\log 2r)}$  time.

**Theorem 2.** *There is a PTAS with ratio  $1 + \frac{1}{r}$  that runs in  $O(n^{2r-1} \times n^{O(\log 2r)})$  time when  $m = O(\log n)$ .*

#### 4. The algorithm for $c \geq O(\log n)$ and $c = O(m)$

In this section, we consider the case where  $c = O(m)$ . We use linear programming and randomized rounding approach. Let  $h_j = (u_j, S_j)$  be a hyperedge. We define  $k_j + 1$  variables,  $x_{j,1}, x_{j,2}, \dots, x_{j,k_j+1}$ .  $x_{j,l} = 1$  indicates that  $P_l^j$  is cut for the embedding of  $h_j$ . For each segment  $P_q^j$  of  $h_j$  and an edge  $e_i^+$  on the ring, we have a constant  $\mu_{i,q,j}$ .  $\mu_{i,q,j} = 1$  if edge  $e_i^+$  is in the segment  $P_q^j$  of  $h_j$ . Otherwise,  $\mu_{i,q,j} = 0$ . We have the following LP formulation.

min  $c$ ;

$$\sum_{l=1}^{k_j+1} x_{j,l} = 1;$$

$$c(e_i^+) = \sum_{j=1}^m \sum_{q=1}^{k_j+1} \mu_{i,q,j} (x_{j,q+1} + x_{j,q+2} + \dots + x_{j,k_j}) \leq c; \quad (1)$$

$$c(e_i^-) = \sum_{j=1}^m \sum_{q=1}^{k_j+1} \mu_{i,q,j} (x_{j,0} + x_{j,1} + \dots + x_{j,q-1}) \leq c. \quad (2)$$

For a fixed hyperedge  $h_j$  and a directed edge  $e_i^+$  ( $e_i^-$ ) on the ring, there is only one  $\mu_{i,q,j}$  for  $q = 0, 1, \dots, k_j$  with value 1. Consider such a segment  $P_q^j$ , where  $e_i^+$  and  $e_i^-$  are in the segment.  $e_i^+$  is used in the embedding of  $h_j$  if one of the segments  $P_{q+1}^j, P_{q+2}^j, \dots, P_{k_j}^j$  is cut for the embedding of  $h_j$ . Likewise,  $e_i^-$  is used in the embedding of  $h_j$  if one of the segments  $P_0^j, P_1^j, \dots, P_{q-1}^j$  is cut for the embedding of  $h_j$ . Therefore, we have (1) and (2).

The fractional version of the linear programming problem can be solved in polynomial time. After we get a fractional solution  $x'_{j,l}$ , independently with probability  $x'_{j,l}$ , we set  $\hat{x}_{j,l} = 1$  and  $\hat{x}_{j,h} = 0$  for the rest of  $h$ . Thus, we obtain an integer solution  $\hat{x}$  for the LP problem. Let  $c_{\text{opt}}$  be the optimal congestion of the LP formulation. Similar to Lemma 3 in [8], we can prove that

**Theorem 3.** *Assume that  $m \geq c_1 \log n$ , and  $c_{\text{opt}} = c_2 \times m$ . Let  $\hat{x}$  be the 0–1 solution obtained by randomized rounding. With probability at least  $1 - n^{-\frac{1}{3}\varepsilon^2 c_2^2 c_1}$ , for any  $e_i^+$  and  $e_i^-$  in  $E_R$ ,*

$$e_i^+(\hat{x}) \leq (1 + \varepsilon)c_{\text{opt}},$$

and

$$e_i^-(\hat{x}) \leq (1 + \varepsilon)c_{\text{opt}}.$$

The proof is given in Appendix A. Using the standard derandomization method for packing integer programs [7], we can have a polynomial deterministic algorithm.

**Theorem 4.** *There is a PTAS for EDHR when  $m \geq O(\log n)$  and  $c_{\text{opt}} \geq O(m)$ .*

#### 5. The general algorithm

The linear programming and randomized rounding approach in Section 4 does not work for the case where  $c_{\text{opt}}$  is small comparing with  $m$ . Here we propose a method that decomposes the set of all hyperedges into two groups so that we can give approximate embeddings using different methods for the two groups.

Consider  $2r$  indices  $i_1, i_2, \dots, i_{2r}$  of edges in  $E_R$ . Let  $e_{i_1}^+, e_{i_2}^+, \dots, e_{i_{2r}}^+$  be  $2r$  edges on the ring. We define

$$R_{i_1, i_2, \dots, i_{2r}} = \{1 \leq j \leq m \mid \text{there exist an } l \text{ such that } e_{i_k}^+ \in P_l^j \text{ for any } k \in \{1, 2, \dots, 2r\}\}.$$

Let  $U_{i_1, i_2, \dots, i_{2r}} = \{1, 2, \dots, m\} - R_{i_1, i_2, \dots, i_{2r}}$ . Let  $x_{\text{opt}}$  be an optimal embedding.  $x_{\text{opt}}|_{R_{i_1, i_2, \dots, i_{2r}}}$  and  $x_{\text{opt}}|_{U_{i_1, i_2, \dots, i_{2r}}}$

denote the reduced embeddings of  $x_{\text{opt}}$  on the sets of hyperedges  $R_{i_1, i_2, \dots, i_{2r}}$  and  $U_{i_1, i_2, \dots, i_{2r}}$ , respectively.

**Lemma 5.**  $|U_{i_1, i_2, \dots, i_{2r}}| \leq 4rc_{\text{opt}}$  and  $|R_{i_1, i_2, \dots, i_{2r}}| \geq m - 4rc_{\text{opt}}$ .

**Proof.** Consider an  $E_j(x_{\text{opt}})$  containing all the  $2r$  edges  $e_{i_1}^+, e_{i_2}^+, \dots, e_{i_{2r}}^+$ . For each edge  $e_{i_k}^+$ , there are at most  $2c_{\text{opt}}$   $H_j$ 's such that  $e_{i_k}^+ \notin E_j(x_{\text{opt}})$ . Thus, there are at most  $4rc_{\text{opt}}$   $H_j$ 's in total with  $e_{i_k}^+ \notin E_j(x_{\text{opt}})$  for some  $i_k$ . Therefore,  $|R_{i_1, i_2, \dots, i_{2r}}| \geq m - 4rc_{\text{opt}}$ .

By definition,  $|U_{i_1, i_2, \dots, i_{2r}}| \leq 4rc_{\text{opt}}$ .  $\square$

Let  $x^{i_1}$  be an embedding of  $H_j$ 's in  $R_{i_1, i_2, \dots, i_{2r}}$  such that every  $H_j$  is cut at edge  $e_{i_1}^+$ . Now, we want to show that

**Lemma 6.** For any fixed index  $1 \leq i_1 \leq n$ , there exist  $2r - 1$  indices  $1 \leq i_2, i_3, \dots, i_{2r} \leq n$  such that for every edge  $e_i^+$  and  $e_i^-$  in  $E_R$ ,

$$e_i^+(x^{i_1}) - e_i^+(x_{\text{opt}}|R_{i_1, i_2, \dots, i_{2r}}) \leq \frac{1}{r}c_{\text{opt}}, \quad \text{and}$$

$$e_i^-(x^{i_1}) - e_i^-(x_{\text{opt}}|R_{i_1, i_2, \dots, i_{2r}}) \leq \frac{1}{r}c_{\text{opt}}.$$

**Proof.** To show the existence of the  $2r - 1$  indices, we give a way to find the  $2r - 1$  indices assuming that  $x_{\text{opt}}$  is known. First, we select an arbitrary edge, say,  $e_1^+ = (1, 2) = (i_1, i_1 + 1)$  on the ring. In the embedding of  $x_{\text{opt}}$ , there are at least  $m - 2c_{\text{opt}}$   $H_j$ 's with  $e_1^+ \in E_j(x_{\text{opt}})$ . That is, there are at most  $2c_{\text{opt}}$   $H_j$ 's with  $e_1^+ \notin E_j(x_{\text{opt}})$ . (If we cut all the  $m$   $H_j$ 's in  $H$  at edge  $e_1^+$ , there are at most  $2c_{\text{opt}}$   $H_j$ 's that are embedded in a way different from that of  $x_{\text{opt}}$ .) Let  $H_r$  be the set of the (at most  $2c_{\text{opt}}$ )  $H_j$ 's with  $e_1^+ \in E_j(x_{\text{opt}})$ .

For every edge  $e_g^+$  on the ring, if there are more than  $c_{\text{opt}}/r$  hyperedges in  $H_r$  with  $e_g^+ \in E_j(x_{\text{opt}})$ , then we select the index  $q$ . Consider the set  $R_{i_1, g}$  of indices.  $j \in R_{i_1, g}$  if  $e_g^+ \in E_j(x_{\text{opt}})$  and  $e_1^+ \in E_j(x_{\text{opt}})$ . If we cut all the (at most  $m - 2c_{\text{opt}}$ )  $H_j$ 's with  $j$  in  $R_{i_1, g}$  on edge  $e_1^+$ , there are at most  $2c_{\text{opt}} - c_{\text{opt}}/r$   $H_j$ 's that are embedded in a way different from that of  $x_{\text{opt}}$ . Set  $H_r = H_r - \{j \mid e_g^+ \in E_j(x_{\text{opt}})\}$  (the set of at most  $2c_{\text{opt}} - c_{\text{opt}}/r$   $H_j$ 's that are embedded in a way different from that of  $x_{\text{opt}}$ ). If the size of  $H_r$  is more than  $c_{\text{opt}}/r$  then we can repeat the process and find another edge  $e_g$  (index). The process continues until the size of  $H_r$  is less than  $c_{\text{opt}}/r$ . The above procedure will stop after at most  $2r - 1$  iterations since each time the size of  $H_r$  is reduced by at least  $c_{\text{opt}}/r$  and the original size of  $H_r$  is at most  $2c_{\text{opt}}$ .

Now, consider any edge  $e_i^+$  with index  $i$  ( $1 \leq i \leq n$ ) not selected in the above procedure. The number of  $H_j$ 's in  $R_{i_1, i_2, \dots, i_{2r}}$  that are not cut correctly at edge  $e_i^+$  in  $x^{i_1}$  is at most the size of  $H_r$ , that is upper bounded by  $c_{\text{opt}}/r$ . Thus, the lemma holds.  $\square$

**Theorem 7.** There is a PTAS for the EDHR problem.

**Proof.** We first compute  $U_{i_1, i_2, \dots, i_{2r}}$  and  $R_{i_1, i_2, \dots, i_{2r}}$ .

*Case 1.*  $|U_{i_1, i_2, \dots, i_{2r}}| \leq C \log n$ . We use the enumerating approach in Section 3 to compute an embedding for the set of hyperedges in  $U_{i_1, i_2, \dots, i_{2r}}$ . For the hyperedges in  $R_{i_1, i_2, \dots, i_{2r}}$ , we simply cut the ring at edge  $e_1^+$ . From Lemma 6 and Theorem 2, the ratio is  $\frac{1}{r} + \frac{1}{r}$ .

*Case 2.*  $|U_{i_1, i_2, \dots, i_{2r}}| > C \log n$ . We use the LP and randomized rounding approach in Section 4 to compute an embedding for the set of hyperedges in  $U_{i_1, i_2, \dots, i_{2r}}$ . For the hyperedges in  $R_{i_1, i_2, \dots, i_{2r}}$ , we simply cut the ring at edge  $e_1^+$ . The LP formulation is as follows:

min  $c$ ;

$k_j + 1$

$$\sum_{l=1}^{k_j+1} x_{j,l} = 1 \quad \text{for } j = 1, 2, \dots, |U_{i_1, i_2, \dots, i_{2r}}|;$$

$$|U_{i_1, i_2, \dots, i_{2r}}| \sum_{j=1}^{k_j+1} \sum_{q=1}^{k_j+1} \mu_{i,q,j} (x_{j,q+1} + x_{j,q+2} + \dots + x_{j,k_j})$$

$$\leq c - c(e_i^+ | R);$$

$$|U_{i_1, i_2, \dots, i_{2r}}| \sum_{j=1}^{k_j+1} \sum_{q=1}^{k_j+1} \mu_{i,q,j} (x_{j,0} + x_{j,1} + \dots + x_{j,q-1})$$

$$\leq c - c(e_i^- | R),$$

where  $c(e_i^+ | R)$  and  $c(e_i^- | R)$  are the number of times that  $e_i^+$  and  $e_i^-$  are used for the embedding of  $H_j$ 's in  $R_{i_1, i_2, \dots, i_{2r}}$ .

Theorem 3 and Lemma 6 ensure that the ratio is  $1 + \varepsilon$  for any  $\varepsilon$ . The standard derandomization approach gives a deterministic algorithm.  $\square$

**Remark.** The NP-hardness of the directed version is still open.

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## Appendix A. The proof of Theorem 3

**Proof.** To prove the theorem, we need the following lemma originally from [7].

**Lemma 8.** Let  $X_1, X_2, \dots, X_n$  be  $n$  independent random 0–1 variables, where  $x_i$  takes 1 with probability  $p_i$ ,  $0 < p_i < 1$ . Let  $X = \sum_{i=1}^n X_i$ , and  $\mu = E[X]$ . Then for  $\delta > 0$ ,  $\Pr(X > \mu + \delta n) < \exp(-\frac{1}{3}n\delta^2)$ .

For a fixed  $i$  and a fixed  $j$ , only one  $\mu_{i,q,j}$  is 1 and the rest are 0. For  $l = 0, 1, \dots, k_j$ , consider such an  $l$  with  $\mu_{i,q,l} = 1$ . For a fixed  $j$ , only one  $x_{j,l}$  is rounded to 1. Thus,  $\mu_{i,q,j}(x_{j,q+1} + x_{j,q+2} + \dots + x_{j,k_j})$  and  $\mu_{i,q,j}(x_{j,0} + x_{j,1} + \dots + x_{j,q-1})$  are also randomly rounded to either 1 or 0 and are independently for different  $j$ 's. Therefore, both

$$c(e_i^+) = \sum_{j=1}^m \sum_{q=1}^{k_j+1} \mu_{i,q,j}(x_{j,q+1} + x_{j,q+2} + \dots + x_{j,k_j})$$

and

$$c(e_i^-) = \sum_{j=1}^m \sum_{q=1}^{k_j+1} \mu_{i,q,j}(x_{j,0} + x_{j,1} + \dots + x_{j,q-1})$$

are sums of  $m$  independent 0–1 random variables. Set

$$\begin{aligned} E[c(e_i^+)] &= \sum_{j=1}^m \sum_{q=1}^{k_j+1} \mu_{i,q,j} E[x_{j,q+1} + x_{j,q+2} + \dots + x_{j,k_j}] \\ &= \mu_i^+ \leq c_{\text{opt}}, \end{aligned}$$

and

$$\begin{aligned} E[c(e_i^-)] &= \sum_{j=1}^m \sum_{q=1}^{k_j+1} \mu_{i,q,j} E[x_{j,0} + x_{j,1} + \dots + x_{j,q-1}] \\ &= \mu_i^- \leq c_{\text{opt}}. \end{aligned}$$

From Lemma 8, for any fixed  $\delta$ ,

$$\Pr(c(e_i^+) > \mu_i^+ + \delta m) \leq \exp\left(-\frac{1}{3}\delta^2 m\right).$$

Consider the set of all clockwise edges  $E_R^+ = \{e_1^+, e_2^+, \dots, e_n^+\}$ ,

$$\Pr(c(e_i^+) > \mu_i^+ + \delta m \text{ for at least one } e_i^+ \in E_R)$$

$$\leq n \times \exp\left(-\frac{1}{3}\delta^2 m\right).$$

Similarly, we can show that

$$\Pr(c(e_i^-) > \mu_i^- + \delta m \text{ for at least one } e_i^- \in E_R)$$

$$\leq n \times \exp\left(-\frac{1}{3}\delta^2 m\right).$$

By assumption,  $m \geq C \log n$ . Thus, we have

$$n \times \exp\left(-\frac{1}{3}\delta^2 m\right) \leq n^{1-\delta^2 C/3}.$$

Therefore, we get a randomized algorithm to find a solution  $x$  for the problem with probability at least  $1 - 2n^{1-\delta^2 C/3}$  such that for any  $e_i^+ \in E_R$  and  $e_i^- \in E_R$ ,  $c(e_i^+) \leq \mu_i + \delta m \leq c_{\text{opt}} + \varepsilon c_{\text{opt}}$ , and  $c(e_i^-) \leq c_{\text{opt}} + \varepsilon c_{\text{opt}}$ , where  $\varepsilon = \frac{\delta}{c}$ .  $\square$

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