

# Constrained pairwise and center-star sequences alignment problems

Yong Zhang<sup>1,3</sup>  $\cdot$  Joseph Wun-Tat Chan<sup>2</sup>  $\cdot$ Francis Y. L. Chin<sup>3</sup>  $\cdot$  Hing-Fung Ting<sup>3</sup>  $\cdot$ Deshi Ye<sup>4</sup>  $\cdot$  Feng Zhang<sup>5</sup>  $\cdot$  Jianyu Shi<sup>6</sup>

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**Abstract** Sequence alignment is a fundamental problem in computational biology, which is also important in theoretical computer science. In this paper, we consider the problem of aligning a set of sequences subject to a given constrained sequence.

⊠ Deshi Ye yedeshi@zju.edu.cn

> Yong Zhang zhangyong@siat.ac.cn

Joseph Wun-Tat Chan cswtchan@gmail.com

Francis Y. L. Chin chin@cs.hku.hk

Hing-Fung Ting hfting@cs.hku.hk

Feng Zhang amyfzhang@gmail.com

Jianyu Shi jianyushi@nwpu.edu.cn

- <sup>1</sup> Shenzhen Institutes of Advanced Technology, Chinese Academy of Sciences, Shenzhen, China
- <sup>2</sup> College of International Education, Hong Kong Baptist University, Kowloon, Hong Kong, China
- <sup>3</sup> Department of Computer Science, The University of Hong Kong, Pok Fu Lam, Hong Kong, China
- <sup>4</sup> College of Computer Science, Zhejiang University, Hangzhou, China
- <sup>5</sup> College of Mathematics and Information Science, Hebei University, Baoding, China
- <sup>6</sup> School of Life Science, Northwestern Polytechnical University, Xi'an, China

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Given two sequences  $A = a_1a_2...a_n$  and  $B = b_1b_2...b_n$  with a given distance function and a constrained sequence  $C = c_1c_2...c_k$ , our goal is to find the optimal sequence alignment of A and B w.r.t. the constraint C. We investigate several variants of this problem. If  $C = c^k$ , i.e., all characters in C are same, the optimal constrained pairwise sequence alignment can be solved in  $O(\min\{kn^2, (t - k)n^2\})$  time, where t is the minimum number of occurrences of character c in A and B. If in the final alignment, the alignment score between any two consecutive constrained characters is upper bounded by some value, which is called GB-CPSA, we give a dynamic programming with the time complexity  $O(kn^4/\log n)$ . For the constrained centerstar sequence alignment (CCSA), we prove that it is NP-hard to achieve the optimal alignment even over the binary alphabet. Furthermore, we show a negative result for CCSA, i.e., there is no polynomial-time algorithm to approximate the CCSA within any constant ratio.

**Keywords** Sequence alignment · Dynamic programming · Complexity

### **1** Introduction

Sequence alignment (Mount 2004) is a fundamental method in computational biology to analyzing the functional, structural, or evolutionary relationships between a set of DNA, RNA, or protein sequences. It is well-defined and had received in-depth study in bioinformatics and theoretical computer science during these years.

Let  $\Sigma$  be the set of alphabets. For *s*, a sequence of *n* characters over  $\Sigma$ ,  $s[x \dots y]$  denotes the substring of *s* from the *x*th character to the *y*th character of *s*, where  $1 \le x < y \le n$ . In particular, let s[x] denote the *x*th character of *s*. In general, the size of the alphabet set is not large. In this paper, we assume that  $|\Sigma| = c$ , where *c* is a constant. E.g., in DNA sequence,  $\Sigma = \{A, C, G, T\}$ .

Given two sequences s and t, the Pairwise Sequence Alignment (PSA) w.r.t. s and t is two sequences s' and t' such that s' and t' have the same length n' and removing all space characters "-" from s' and t' gives s and t respectively. Let function  $\alpha$ define the alignment positions, i.e.,  $\alpha(s, i)$  denotes the position of the character in s' corresponds to character s[i]. So  $s'[\alpha(s, i)] = s[i]$ . For a given distance function  $\delta(x, y) \geq 0$  which measures the mutation distance between two characters, where  $x, y \in \Sigma \cup \{-\}$ , the pair-wise score of two length n' sequences s' and t' is defined as  $\sum_{1 \le x \le n'} \delta(s'[x], t'[x])$ . Usually,  $\delta(x, x) = 0$  for any  $x \in \Sigma$ . The optimal pairwise sequence alignment with the minimum alignment score can be solved in  $O(n^2)$  time by dynamic programming, where *n* is the length of the longer sequence. The problem of finding the longest common subsequence (LCS) of two sequences is a special case of the sequence alignment, which is well-studied for many years. Note that the target of the LCS problem is to find the maximized length of a common subsequence of two strings, however, this maximization problem can be converted to the minimization problem by defining the distance function  $\delta(\cdot, \cdot)$ , such that  $\delta(x, -) = 1$ ,  $\delta(-, y) = 1$ ,  $\delta(x, y) = 2$  if  $x \neq y$  and  $\delta(x, y) = 0$  if x = y, where both x and y are characters in these two strings and not '-'. The LCS problem can be solved in  $O(n^2)$  time (Cormen et al. 2009). The only subquadratic algorithm for LCS was given by Masek

and Paterson (1980), which runs in  $O(n^2/\log n)$  if the size of the alphabet set is bounded by a constant. Iliopoulos and Sohel Rahman (2008) introduced the LCS problem with fixed gap (FIG), in which the distance between any two consecutive characters in the common subsequence in both input sequences are bounded by a given parameter K. They presented an  $O(n^2 + R \log \log n)$  algorithm, where R is the total number of matches between the input sequences.

In the *multiple sequence alignment* (MSA) problem, we are given *m* sequences  $S = \{s_1, s_2, \ldots, s_m\}$  with maximum length *n*. The output of an MSA is an alignment matrix *A*, with *m* rows and  $n'(\ge n)$  columns of characters over  $\Sigma \cup \{-\}$ , such that removing space characters from the *i*th row of *A* gives  $s_i$  for  $1 \le i \le m$ . The sum-ofpairs (SP) score of the MSA *A* is defined to be the sum of the pair-wise scores of all pairs of the sequences, i.e.,

$$score(A) = \sum_{1 \le i < j \le m} \sum_{1 \le p \le n'} \delta(A_{i,p}, A_{j,p})$$

where  $A_{i,p}$  and  $A_{j,p}$  are the characters at the *i*- and *j*th row and the *p*th column of *A*, respectively. Star alignment is also an important alignment in computational biology (Setubal and Meidanis 1997). In the star alignment, a center sequence *s* shall be identified, and it is used as the "center of star" when aligning with all other sequences. A sequence is a center if it is the most similar to all the rest using pairwise alignment. It was shown in Bonizzoni and Vedova (2001) and Wang and Jiang (1994) that finding an alignment matrix with the minimum sum-of-pair alignment score for  $m \ge 3$  sequences is NP-Hard. There are a number of heuristics which approximate the optimal alignment, some with guaranteed worst case approximation ratio, and some with good performance in practice, e.g., BLAST (1990), Clustal W (2007). One of the approximate the optimal alignment within a factor of 2 - 2/m in  $O(mn^2)$  time, where *m* is the number of sequences.

According to the biology knowledge, some residues in the multiple sequence must be aligned in the same columns. For example, His12, Lys41 and His119 should be aligned in the same columns when consider the alignment of RNase sequence. That motivates to study the *constrained sequence alignment problem*, which was introduced by Tang et al. (2003). In the *constrained multiple sequence alignment problem* (CMSA), we are given, in addition to the inputs of the multiple sequence alignment problem, a constrained sequence  $C = c_1 c_2 \dots c_k$ , where C is a common subsequence of all  $s_i \in \{s_1, s_2, \dots, s_m\}$ . The solution of a CMSA problem is a constrained alignment matrix A which is an alignment matrix such that each character in C appears in an entire column of A and also in the same order, i.e. there exists a list of integers  $\{g_1, g_2, \dots, g_k\}$  where  $1 \leq g_1 < \dots < g_k \leq n'$  and for all  $1 \leq i \leq m$  and for all  $1 \leq j \leq k$ , we have  $A_{i,g_i} = c_j$ .

Let *A* be a CMSA matrix for  $S = \{s_1, s_2, ..., s_m\}$  and the constrained sequence *C*. Define *score*(*A*) to be the score of the CMSA matrix *A*. Let  $A_{S,C}^*$  be the optimal CMSA matrix and  $A_{S,C}$  be the CMSA matrix derived by an approximation algorithm. The approximation algorithm is said to have an approximation ratio *r* if and only if for any *S* and *C*,

$$\frac{score(A_{S,C})}{score(A_{S,C}^*)} \le r.$$

For two sequences of a CMSA, it is called *constraint pairwise sequence alignment* (CPSA), Tang et al. (2003) gave an algorithm with both the time and space complexity  $O(kn^4)$ , where k is the length of the constrained sequence. The result was improved by Chin et al. (2004) to  $O(kn^2)$ . To approximate the optimal constrained alignment for  $m \ge 3$  sequences, Chin et al. (2005) introduced the *constrained center-star sequence alignment* (CCSA). Similar to the center-star sequence alignment to approximate the multiple sequences alignment, the optimal constrained center-star sequence alignment has an approximation ratio of no more than 2 - 2/m. However, the proposed algorithm for finding the optimal constrained center-star sequence alignment (CCSA) for m sequences takes  $O(Cmn^2)$  time, where C is the total number of occurrences of the constrained sequence in the m sequences. As C can be exponential, the time complexity for CCSA can be exponential.

In this paper, we consider various constrained sequence alignment problems. Note that *the constrained pairwise sequence alignment* can be solved in  $O(kn^2)$  Chin et al. (2004). In this work, we give an  $O((t - k)n^2)$  algorithm to solve the CPSA problem in which the constrained sequence  $C = c^k$ , and t is the minimum number of occurrences of character c in the two sequences. Combined with the previous result, this variant can be solved in  $O(\min\{kn^2, (t - k)n^2\})$  time when k < t and  $O(n^2)$  when k = t. Next, we study another variant of the constrained pairwise sequence alignment, say *gap-bounded constrained pairwise sequence alignment* (GB-CPSA), in which the alignment score between any two consecutive constrained characters is upper bounded by some value in the final alignment. For the GB-CPSA, we give a dynamic programming with the time complexity  $O(kn^2 + \sum_{p=1}^{k} R_{p-1}R_pn^2/\log n) = O(kn^4/\log n)$ , where  $R_p$  is the number of matches on  $c_p$  in A and B. Finally, we show that *the constrained center-star sequence alignment* problem for multiple sequences is NP-hard even if the size of the alphabet set is two, and this problem cannot be approximated within a constant factor for some distance function.

In the remainder of this paper, we first consider two variants of the constrained pairwise sequence alignment in Sect. 2. The we turn to study the complexity of the constrained center-star sequence alignment problem in Sect. 3. The conclusion is given in Sect. 4.

# 2 Constrained pairwise sequence alignment

In this section we study the constrained pairwise sequence alignment (CPSA) problem. Given two sequences  $A = a_1a_2...a_n$  and  $B = b_1b_2...b_n$ , a constrained sequence  $C = c_1c_2...c_k$ , and a distance function  $\delta$ , the CPSA problem is to find the minimumscore PSA, A' and B', such that A' and B' have the same length  $n' \ge n$  and  $A'[g_i] = B'[g_i] = c_i$  for  $1 \le i \le k$  and some integers  $1 \le g_1 < g_2 < \cdots < g_k \le n'$ . The distance function  $\delta(x, y)$  is a score of aligning the character x and the character y, where  $x, y \in \Sigma \cup \{-\}$ . The distance function  $\delta$  always satisfies  $\delta(x, x) = 0$  for any  $x \in \Sigma \cup \{-\}$ . Let S(i, j, p) denote the optimal CPSA score for the sequences A[1...i] and B[1...j] with the constrained sequence C[1...p] under the distance function  $\delta$ . The function S(i, j, p) can be described recursively as follows as in Chin et al. (2004).

$$S(i, j, p) = \min \begin{cases} S(i - 1, j - 1, p - 1) & \text{if } a_i = b_j = c_p \\ S(i - 1, j - 1, p) + \delta(a_i, b_j) & \text{if } i, j > 0 \\ S(i - 1, j, p) + \delta(a_i, -) & \text{if } i > 0 \\ S(i, j - 1, p) + \delta(-, b_j) & \text{if } j > 0 \end{cases}$$

For the boundary cases, S(i, 0, 0) = S(0, j, 0) = 0 and  $S(0, j, \ell) = S(i, 0, \ell) = \infty$ for  $0 \le i, j \le n$  and  $0 < \ell \le k$ . The last equation means there is no way to align an empty sequence with a non-empty constraint sequence.

This algorithm computes all the entries S(i, j, p) in  $O(kn^2)$  time because  $i, j \le n$ and  $p \le k$  and each entry can be computed in constant time.

# 2.1 Constrained sequence $C = c^k$

In this part, we consider a special case where  $c_1 = c_2 = \ldots = c_k$ , i.e., the *k* characters are all the same. Let that character be *c* and denote  $C = c^k$ . Without loss of generality, assume that the number of occurrences of character *c* in *B* is *t* and it is no more than the number of occurrences of *c* in *A*. Consider the case when *k* approaches *t*, previous algorithm (Chin et al. 2004) takes  $O(tn^2)$  time. In the following we give a more efficient algorithm when *k* is close to *t*. Our algorithm takes  $O(n^2)$  time when k = t and  $O((t - k)n^2)$  time when k < t.

Now we show how to find the optimal PSA score in  $O((t-k)n^2)$  time. Let T(i, j, q) denote the optimal PSA score for sequences  $A[1 \dots i]$  and  $B[1 \dots j]$  and the distance function  $\delta$  such that in the PSA, A' and B', there are exactly q occurrences of character c in B' matched with a character (in A') other than c. Precisely, there are exactly q integers  $g_1, g_2, \dots, g_q$  with  $B'[g_i] = c$  and  $A'[g_i] \neq c$ . The function T(i, j, q) can be defined recursively as follows:

$$T(i, j, q) = \min \begin{cases} T(i - 1, j - 1, q) + \delta(a_i, b_j) & \text{if } b_j \neq c \text{ or } a_i = b_j = c \\ T(i - 1, j - 1, q - 1) + \delta(a_i, b_j) & \text{if } b_j = c \text{ and } a_i \neq c \\ T(i - 1, j, q) + \delta(a_i, -) & \text{if } i > 0 \\ T(i, j - 1, q - 1) + \delta(-, c) & \text{if } b_j = c \\ T(i, j - 1, q) + \delta(-, b_j) & \text{if } b_j \neq c \end{cases}$$

In the recursive step, there are three choices for the optimal PSA to proceed. (Note that the first and the second cases, the fourth and the fifth cases are mutually exclusive, respectively.) The optimal PSA could either (i) match  $a_i$  with  $b_j$  (ii) match  $a_i$  with space, or (iii) match  $b_j$  with space. Obviously, the optimal PSA should yield the minimum score among these three choices. For (i), the PSA would include the distance between  $a_i$  and  $b_j$  and proceed with shortened inputs  $A[1 \dots i-1]$  and  $B[1 \dots j-1]$ . In case of  $b_j \neq c$ , or  $a_i = b_j = c$ , there are exactly q occurrences (otherwise q - 1) of character c in  $B[1 \dots j-1]$  will match with a character other than c. For (ii), the

PSA would include the distance between  $a_i$  and the space character "-". Since only  $a_i$  is matched but not  $b_j$ , the PSA proceeds with  $A[1 \dots i - 1]$  and  $B[1 \dots j]$  and it still requires exactly q occurrences of character c in  $B[1 \dots j]$  matched with a character other than c. For (iii), the PSA would include the distance between  $b_j$  and the space character "-" and proceed with  $A[1 \dots i]$  and  $B[1 \dots j - 1]$ . However, it differs for the case where  $b_j = c$  and  $b_j \neq c$ . If  $b_j = c$ , we have this character c matched with a character "-", which is not c. Thus the subsequent PSA requires only q - 1 occurrences of character c in  $B[1 \dots j - 1]$  matched with a character other than c. Otherwise, it still requires exactly q occurrences of character c in  $B[1 \dots j - 1]$  matched with a character other than c.

Remind that in this section our target is to find the CPSA with constrained sequence  $c^k$ . In fact, it is equivalent to find the minimum-score PSA among the PSA with exactly *i* occurrences of character *c* in *B* matched with a character other than *c* for i = 0 to t - k. Therefore, the required CPSA is the minimum-score PSA among those correspond to T(m, n, i) for  $0 \le i \le t - k$ . In our algorithm we can compute the entries T(i, j, q) for all  $i, j \le n, q \le t - k$ . Since each entry can be computed in constant time, the algorithm can find CPSA in  $O((t - k)n^2)$  time.

Combining the algorithm of Chin et al. (2004) with time complexity  $O(kn^2)$  and our new algorithm with time complexity  $O((t - k)n^2)$  if k < t and  $O(n^2)$  if k = t, we can compute the required CPSA efficiently as shown in the following theorem.

**Theorem 1** The CPSA problem for two sequences with length n, the constrained sequence of  $c^k$ , and a given distance function can be solved in  $O(\min\{k, t - k\}n^2)$  if k < t and in  $O(n^2)$  if k = t where t is the minimum number of occurrences of character c in these two sequences.

#### 2.2 Gap-bounded constrained pairwise sequence alignment

In this part, we consider the gap-bounded constrained pairwise sequence alignment (GB-CPSA), i.e., given two sequences  $A = a_1a_2...a_n$ ,  $B = b_1b_2...b_n$ , the constrained sequence  $C = c_1c_2...c_k$  and a distance function  $\delta$ , the GB-CPSA is to find a minimum-score PSA, A' and B', such that A' and B' have the same length  $n' \ge n$ ,  $A'[g_i] = B'[g_i] = c_i$  for some integers  $1 \le g_1 < g_2 < \cdots < g_k \le n'$  and the alignment score between  $A'[g_i + 1...g_{i+1} - 1]$  and  $B'[g_i + 1...g_{i+1} - 1]$  is upper bounded by K. Roughly speaking, any two consecutive characters in the constrained sequence are not far away in the final alignment. Different from FIG (Iliopoulos and Sohel Rahman 2008), our problem focuses on the bounded gap of the consecutive characters in the constrained sequence, but not in the common subsequence. Between two consecutive constrained characters in the final alignment, some common characters may be aligned together. Due to the distribution of the constrained characters in the input sequences A and B, there may not exist a PSA to satisfy the gap-bounded property. In this case, the algorithm should output "No alignment".

Let S(s, t) be the minimum alignment score w.r.t. sequences s and t. Let  $S_1(i, j, p)$  be the minimum score of GB-CPSA w.r.t. A[1...i], B[1...j] and the constrained sequence C[1...p] such that the constrained character  $c_p$  is aligned with  $a_i$  and  $b_j$ .  $S_2(i, j, p)$  is defined similarly but the constrained character  $c_p$  is not aligned with  $a_i$ 

and  $b_j$ . Note that  $S_1(i, j, 0)$  and  $S_2(i, j, 0)$  are the traditional alignment score without considering the constrained sequence. In details,  $S_1(i, j, p)$  and  $S_2(i, j, p)$  are given as follows.

- $S_1(i, j, p)$ 
  - If  $a_i \neq c_p$  or  $b_j \neq c_p$ ,  $S_1(i, j, p) = +\infty$ .
  - If  $a_i = b_j = c_p$ , the value of  $S_1(i, j, p)$  depends on the previous values  $S_1(i', j', p-1)$  and the alignment between A[i'+1...i-1] and B[j'+1...j-1] such that the alignment score between  $c_{p-1}$  and  $c_p$  is bounded by *K*. Formally,

$$S_1(i, j, p) = \min_{i', j'} \{S_1(i', j', p-1) + S(A[i'+1...i-1], B[j'+1...j-1])\} + \delta(c_p, c_p)$$

where  $a_{i'} = b_{j'} = c_{p-1}$  and  $S(A[i' + 1 \dots i - 1], B[j' + 1 \dots j - 1]) \le K$ . •  $S_2(i, j, p)$ 

Since the constrained character  $c_p$  is aligned before  $a_i$  and  $b_j$ , the score function follows the traditional computation for sequence alignment, i.e.,

$$S_{2}(i, j, p) = \min \begin{cases} \min\{S_{1}(i-1, j-1, p), S_{2}(i-1, j-1, p)\} + \delta(a_{i}, b_{j}) & \text{if } i, j > 0\\ \min\{S_{1}(i-1, j, p), S_{2}(i-1, j, p)\} + \delta(a_{i}, -) & \text{if } i > 0\\ \min\{S_{1}(i, j-1, p), S_{2}(i, j-1, p)\} + \delta(-, b_{j}) & \text{if } j > 0 \end{cases}$$

After all entries of  $S_1(i, j, p)$  and  $S_2(i, j, p)$  are computed for  $1 \le i, j \le n$  and  $1 \le p \le k$ , the minimum score of the GB-CPSA is

$$\min\{S_1(n, n, k), S_2(n, n, k)\}.$$

There are  $kn^2$  entries in  $S_1(i, j, p)$  and  $S_2(i, j, p)$ , respectively. The recursive formula of  $S_1(i, j, p)$ ,  $S_2(i, j, p)$  can be determined in constant time given previous values of  $S_1()$  and  $S_2()$ . To compute the value of  $S_1(i, j, p)$ , the previous alignments on  $c_{p-1}$  with the bounded gap to  $c_p$  have to be considered. Thus, we must do the following two jobs:

- (1) efficiently find the alignments such that of  $a_{i'} = b_{j'} = c_{p-1}$ , and
- (2) efficiently compute the alignment score between A[i' + 1...i 1] and B[j' + 1...j 1].

For the first job, we use two arrays to store the sorted positions of each character  $c_p$  in A and B, respectively. By scanning the sequences A and B once, the arrays for  $c_p$   $(1 \le p \le k)$  can be constructed in O(n). When computing  $S_1(i, j, p)$ , we should determine the positions i' and j' such that  $a_{i'} = b_{j'} = c_{p-1}$ . Given i and j, the largest i' and j' with i' < i, j' < j and  $a_{i'} = b_{j'} = c_{p-1}$  can be achieved in  $O(\log n)$  time by binary search. Whereas  $(a_{i'}, b_{j'})$  is the possible closest match satisfying the constrained character  $c_{p-1}$  w.r.t.  $S_1(i, j, p)$ . Each of the previous matches can be found in O(1) time by tracing the arrays for  $c_{p-1}$  in A and B.

For the second job, the  $O(n^2/\log n)$  time algorithm from Masek and Paterson (1980) can be applied to our problem because we do not need to consider the constrained characters and this job is computing the alignment score of two strings with a constant size of alphabets set. The value of  $S(A[i'+1...i-1], B[j'+1...j-1], \phi)$  can be achieved in  $O((i - i')(j - j')/\log n)$  time, which is upper bounded by  $O(n^2/\log n)$ .

Let  $R_p$  be the total number of matches on  $c_p$  in A and B. Thus, given previous values,  $S_1(i, j, p)$  can be computed in

$$O(\log n + R_{p-1}n^2/\log n) = O(R_{p-1}n^2/\log n).$$

For  $c_p$ , there are  $n^2$  entries of  $S_1(i, j, p)$ , besides these entries such that  $a_i = b_j = c_p$ , the values of all other entries can be determined to be  $+\infty$  in constant time. Therefore, the complexity to fill all  $S_1(i, j, p)$  entries w.r.t.  $c_p$  is

$$O(R_{p-1}R_pn^2/\log n).$$

Considering all constrained characters in  $C = c_1 c_2 \dots c_k$ , filling all entries of  $S_1(i, j, p)$  takes time

$$O\left(\sum_{p=1}^{k} R_{p-1} R_p n^2 / \log n\right).$$

From above analysis, we conclude the following theorem.

**Theorem 2** The Gap-Bounded Constrained Pairwise Sequence Alignment (GB-CPSA) can be solved in  $O(kn^4/\log n)$ .

*Proof* The time complexity to compute all entries of  $S_1(i, j, p)$  is  $O\left(\sum_{p=1}^k R_{p-1}R_p n^2/\log n\right)$ . The time complexity to compute all entries of  $S_2(i, j, p)$  is  $O(kn^2)$ . Determine the minimum score takes O(1) time by comparing  $S_1(n, n, k)$  and  $S_2(n, n, k)$ . Thus, the time complexity for solving GB-CPSA is

$$O\left(kn^{2} + \sum_{p=1}^{k} R_{p-1}R_{p}n^{2}/\log n\right) = O(kn^{4}/\log n).$$

*Remark* When compute each entry of  $S_1(i, j, p)$  and  $S_2(i, j, p)$ , we may use two tables  $T_1(i, j, k)$  and  $T_2(i, j, k)$  to store the previous entry leads to the minimum value. Thus, for the GB-CPSA problem, the sequence alignment with the minimum score can be achieved in linear time by tracing back from  $T_1(i, j, p)$  and  $T_2(i, j, p)$ .

# 3 Constrained center-star sequence alignment

In this section, we study the problem of *constrained center-star sequence alignment* (*CCSA*). This problem is similar to the CMSA problem since the goal is also to find an optimal alignment matrix with a constrained sequence. However, CCSA has a different way of calculating the score of an alignment matrix. In CMSA, we minimize the sum of all pairwise scores of the alignment matrix. In CCSA, we minimize, for  $s \in S$ , the sum of the pairwise scores of s with every sequence in  $S - \{s\}$ . Precisely speaking, the score of an alignment matrix A in CCSA is defined to be

$$\min_{1 \le i \le m} \sum_{1 \le j \le m, j \ne i} \sum_{1 \le p \le n'} \delta(A_{i,p}, A_{j,p}).$$

For an optimal alignment matrix A, we call a particular sequence  $s_{\ell}$  the *center sequence* if the score of A is

$$\sum_{1 \le j \le m, j \ne \ell} \sum_{1 \le p \le n'} \delta(A_{\ell,p}, A_{j,p}).$$

## 3.1 NP-hardness of CCSA

In this part, we prove that CCSA is NP-Hard over binary alphabet, i.e., the size of alphabet set is 2. To prove this, we give a polynomial time reduction from the NP-hard problem *Maximum Independent Set (MIS)* (Garey and Johnson 1979) to CCSA. Consider the decision version of MIS as follows.

Maximum Independent Set (MIS): Given a graph G = (V, E) and an integer k, is there a subset  $V' \subseteq V$  of vertices for |V'| = k such that each edge in E is incident on at most one vertex in V'?

For any instance of the decision version of MIS which includes a graph G = (V, E)and an integer k, we give a polynomial-time transformation to an instance of CCSA which includes the alphabet set  $\Sigma$ , a set of sequences S over  $\Sigma$ , a constrained sequence C, and a distance function  $\delta$ . We define the transformation as  $\Pi$ , i.e.,  $\Pi(G, k) = (\Sigma, S, C)$ . Let  $V = \{v_1, v_2, \ldots, v_n\}$  and  $E = \{e_1, \ldots, e_m\}$ . The transformation  $\Pi(G, k)$ , constructed according to G and k, is shown as follows. We have  $\Sigma = \{a, b\}$  and  $S = \{t_1, \ldots, t_m, s_1, \ldots, s_m\}$ , where  $t_1 = \cdots = t_m = (aba)^n$  and each  $s_i$  is an encoding corresponding to an edge  $e_i$ . Suppose  $e_i = (v_p, v_q)$  with p < q, then  $s_i = (bab)^{p-1}(baa)(bab)^{q-p-1}(aab)(bab)^{n-q}$ . The constrained sequence  $c = b^k$ . The distance function  $\delta$  between any two characters in  $\Sigma \cup \{-\}$  of the constructed CCSA instance is defined as follows.

	a	b	-
a	0	1	2
b	1	0	2
-	2	2	0

Consider  $s_i = (bab)^{p-1}(baa)(bab)^{q-p-1}(aab)(bab)^{n-q}$  and  $t_j = (aba)^n$  for  $1 \le i, j \le m$  and  $e_i = (v_p, v_q)$  where p < q. Two possible alignments of  $s_i$  and  $t_j$  are

$$s'_{i} = -(bab)^{p-1}(baa)(bab)^{q-p-1}(aab)(bab)^{n-q}$$
  
$$t'_{i} = (aba)^{n} -$$
(1)

and

$$s'_{i} = (bab)^{p-1}(baa)(bab)^{q-p-1}(aab)(bab)^{n-q} - t'_{j} = -(aba)^{n}.$$
(2)

It is easy to check that both alignments yield a score of n + 3. The following lemma shows that except the above two alignments all other alignments of  $s_i$  and  $t_i$  yield a score greater than n + 3.

**Lemma 1** For any  $1 \le i, j \le m$ , the alignments of  $s_i$  and  $t_j$  according to Alignments (1) and (2) yield a score of n + 3. All other alignments of  $s_i$  and  $t_j$  yield a score greater than n + 3.

*Proof* Suppose edge  $e_i = (v_p, v_q)$ , then  $s_i = (bab)^{p-1}(baa)(bab)^{q-p-1}(aab)$  $(bab)^{n-q}$ . We also have  $t_i = (aba)^n$ .

We consider all possible alignments of  $s_i$  and  $t_j$  and divide them into 7 cases. These cases are characterized by how the sub-string "baa" in  $s_i$  "overlaps" with the *p*th "aba" in  $t_j$ . The positions of the characters of sub-string "baa" in  $s_i$  (as well as the *p*th "aba" in  $t_j$ ) are 3p-2, 3p-1 and 3p, respectively. The following cases consider all possible relationships between the values of  $\alpha(s, 3p - 2), \alpha(s, 3p - 1), \alpha(s, 3p), \alpha(t, 3p - 2), \alpha(t, 3p - 1), and \alpha(t, 3p)$ .

**Case (1)**  $\alpha(t, 3p - 2) > \alpha(s, 3p)$ .

In this case the corresponding sub-string "baa" in  $s_i$  appears before the *p*th "aba" in  $t_j$  in the alignment. We further consider the sub-case  $\alpha(s, 3p) < \alpha(t, 3p - 2) \le \alpha(s, 3p + 1)$ . The other sub-case  $\alpha(s, 3p + 1) < \alpha(t, 3p - 2)$  can be proved similarly. The alignment can be seen as breaking each of  $s_i$  and  $t_j$  into two parts and aligning corresponding parts of  $s_i$  and  $t_j$  as follows.

si	$(bab)^{p-1}$ baa	$(bab)^{q-p-1}(aab)(bab)^{n-q}$
$t_j$	$(aba)^{p-1}$	$\mathbf{aba}(aba)^{n-p}$

The first part of  $s_i$ , denote by  $s_i^1$  has p + 1 "a" and 2p - 1 "b", the first part of  $t_j$ , denote by  $t_j^1$  has 2p - 2 "a" and p - 1 "b". Even if all "a" in  $s_i^1$  and all "b" in  $t_j^1$  are matched, there are still p - 3 "a" and p "b" unmatched. Therefore, the score of the first part of this alignment is at least p + 3 because if all unmatched p - 3 "a" align with p - 3 "b" there are still 3 "b" that must align with space characters. For the same reason, the score of second part of this alignment is at least n - p + 4. Hence, the total score of this pairwise alignment is at least n + 7.

**Case** (2)  $\alpha(s, 3p - 1) < \alpha(t, 3p - 2) \le \alpha(s, 3p)$ . The alignment can be seen as breaking each of  $s_i$  and  $t_j$  into two parts and aligning corresponding parts of  $s_i$  and  $t_j$  as follows.

<i>s</i> <sub>i</sub>	$(bab)^{p-1}$ ba	$\mathbf{a}(bab)^{q-p-1}(aab)(bab)^{n-q}$
$t_j$	$(aba)^{p-1}$	$aba(aba)^{n-p}$

Similarly to the counting in Case (1), we have the score of the first part of the alignment at least p + 2 and the score of the second part of the alignment at least n - p + 2. Hence, the total score of the alignment is at least n + 4.

**Case (3)**  $\alpha(s, 3p - 2) < \alpha(t, 3p - 2) \le \alpha(s, 3p - 1)$ . The alignment can be seen as breaking each of  $s_i$  and  $t_j$  into two parts and aligning corresponding parts of  $s_i$  and  $t_j$  as follows.

s <sub>i</sub>	$(bab)^{p-1}\mathbf{b}$	$aa(bab)^{q-p-1}(aab)(bab)^{n-q}$
$t_j$	$(aba)^{p-1}$	$\mathbf{aba}(aba)^{n-p}$

In the first part, the minimal alignment score is p + 1, and this happens only when  $s_i^{1'} = (bab)^{p-1}b$  and  $t_j^{1'} = -(aba)^{p-1}$ . Otherwise, the score is greater than p+1. For the second part, consider the sub-string of "aa", the (n - p - 1) sub-strings of "bab", and the sub-string of "aab", which form  $s_i^2$ . The score of each of "aa" and the (n-p-1) "bab" is at least 1. Moreover, there must be at least one space character to be inserted to  $s_i^2$  for the alignment with  $t_j^2$ . So the minimum score is n - p + 2 and it happens only when  $s_i^{2'} = aa(bab)^{q-p-1}(aab)(bab)^{n-q}$  and  $t_j^{2'} = (aba)^{n-p+1}$ . Thus the minimum score for aligning  $s_i$  and  $t_j$  in this case is n + 3 and it happens only when  $s_i' = (bab)^{p-1}(baa)(bab)^{q-p-1}(aab)(bab)^{n-q}$  and  $t_j' = -(aba)^n$ . Otherwise, the score is greater than n + 3.

**Case** (4)  $\alpha(t, 3p-2) \le \alpha(s, 3p-2)$  and  $\alpha(t, 3p) \ge \alpha(s, 3p)$ . We further consider the sub-case  $\alpha(s, 3p-3) < \alpha(t, 3p-2) \le \alpha(s, 3p-2)$  and  $\alpha(s, 3p) \le \alpha(t, 3p) < \alpha(s, 3p+1)$ . The other sub-cases can be proved similarly. The alignment can be seen as breaking each of  $s_i$  and  $t_j$  into three parts and aligning corresponding parts of  $s_i$ and  $t_j$  as follows.

ç.	$(bab)^{p-1}$	baa	$(bab)^{q-p-1}(aab)(bab)^{n-q}$
s <sub>i</sub>	$(DuD)^{i}$	Daa	$(bub)^{I}$ $(uub)(bub)$ $I$
$t_j$	$(aba)^{p-1}$	aba	$(aba)^{n-p}$

The alignment score for the first part is at least p + 3, for the second part is at least 2, and for the third part is at least n - p + 2. Thus the alignment score of  $s_i$  and  $t_j$  is at least n + 7.

**Case (5)**  $\alpha(t, 3p - 2) \le \alpha(s, 3p - 2)$  and  $\alpha(s, 3p - 1) \le \alpha(t, 3p) \le \alpha(s, 3p)$ . The alignment can be seen as breaking each of  $s_i$  and  $t_j$  into two parts and aligning corresponding parts of  $s_i$  and  $t_j$  as follows.

si	$(bab)^{p-1}$ ba	$\mathbf{a}(bab)^{q-p-1}(aab)(bab)^{n-q}$
$t_j$	$(aba)^{p-1}$ aba	$(aba)^{n-p}$

Similarly to that of Case (3), in the first part, the minimal alignment score is p + 1, and this happens only when  $s_i^{1'} = -(bab)^{p-1}ba$  and  $t_j^{1'} = (aba)^{p-1}$ . Otherwise, the score is greater than p + 1. For the second part, the minimum alignment score is n - p + 2 and it happens only when  $s_i^{2'} = a(bab)^{q-p-1}(aab)(bab)^{n-q}$  and  $t_j^{2'} = (aba)^{n-p}$ . Thus the minimum score for aligning  $s_i$  and  $t_j$  in this case is n + 3 and it happens only when  $s_i' = -(bab)^{p-1}(baa)(bab)^{q-p-1}(aab)(bab)^{n-q}$  and  $t_j' = (aba)^n$ . Otherwise, the score is greater than n + 3.

**Case** (6)  $\alpha(t, 3p-2) \le \alpha(s, 3p-2)$  and  $\alpha(s, 3p-2) \le \alpha(t, 3p) \le \alpha(s, 3p-1)$ . The alignment can be seen as breaking each of  $s_i$  and  $t_j$  into two parts and aligning corresponding parts of  $s_i$  and  $t_j$  as follows.

Si	$(bab)^{p-1}\mathbf{b}$	$aa(bab)^{q-p-1}(aab)(bab)^{n-q}$
tj	$(aba)^{p-1}$ aba	$(aba)^{n-p}$

Similar to that in Case (2), the total score of this alignment is at least n + 4.

**Case** (7)  $\alpha(t, 3p) < \alpha(s, 3p - 2)$ . The alignment can be seen as breaking each of  $s_i$  and  $t_j$  into two parts and aligning corresponding parts of  $s_i$  and  $t_j$  as follows.

Si	$(bab)^{p-1}$	$\mathbf{baa}(bab)^{q-p-1}(aab)(bab)^{n-q}$
$t_j$	$(aba)^{p-1}$ aba	$(aba)^{n-p}$

Similar to that in Case (1), the total score of this alignment is at least n + 7.

**Corollary 1** If the score for the constrained center-star sequence alignment is at most m(n + 3), the center sequence must be one of  $t_j$  for  $1 \le j \le n$ .

*Proof* By Lemma 1, the alignment score between  $s_i$  and  $t_j$  for any i and j is at least n + 3. It is obvious that the alignment score between  $s_i$  and  $s_j$  for any  $i \neq j$  is more than 0. Thus, if  $s_i$  for some i is the center sequence, the total alignment score must be greater than m(n + 3). Therefore, we can assume that the center sequence must be one of  $t_i$  for  $1 \leq j \leq n$ .

**Lemma 2** There is an independent set for G of size k if and only if there is a constrained center sequence alignment for the transformed instance  $\Pi(G, k)$  of CCSA with score m(n + 3).

*Proof* First, if *G* has an independent set  $\Phi$  of size *k*, we give a constrained center sequence alignment with score m(n + 3). The alignment has  $t_1$  (or any one of  $t_j$  for  $1 \le j \le m$ ) as the center sequence. We have  $t'_j = -(aba)^n - \text{ for } 1 \le j \le n$ . For a sequence  $s_i$  corresponding to edge  $e_i = (v_p, v_q)$  with p < q, we have  $s'_i$  as follows.

• If 
$$v_p \in \Phi$$
,  $s'_i = -s_i = -(bab)^{p-1}(baa)(bab)^{q-p-1}(aab)(bab)^{n-q}$ ,

• If  $v_p \notin \Phi$ ,  $s_i' = s_i - (bab)^{p-1}(baa)(bab)^{q-p-1}(aab)(bab)^{n-q}$ .

Recall that the constrained sequence is  $b^k$ . We prove that the constrain is satisfied in the alignment. Consider a vertex  $v_x \in \Phi$ . We can see that  $t'_j[3x] = b$ for  $1 \le j \le m$ . For an edge  $e_i = (v_x, v_y)$  incident to  $v_x$  and if x < y, then  $s'_i = -(bab)^{x-1}(baa)(bab)^{y-x-1}(aab)(bab)^{n-y}$  and  $s'_i[3x] = b$ . If y < x, since  $v_y \notin \Phi$ ,  $s'_i = (bab)^{y-1}(baa)(bab)^{x-y-1}(aab)(bab)^{n-x}$ - and we also have  $s'_i[3x] = b$ . If an edge  $e_i$  is not incident to  $v_x$ , we have  $s'_i[3x] = b$ . Therefore, the constrain is satisfied.

For the alignment score, by Lemma 1, we can see that the score of the alignment  $t'_1$  and  $s'_i$  for each  $1 \le i \le m$  is n + 3 and the score of the alignment  $t'_1$  and  $t'_j$  for each  $1 \le j \le m$  is 0. Thus the total score is m(n + 3).

Second, if there is a constrained center sequence alignment with score m(n+3), we prove that there is an independent set  $\Phi$  for *G* of size *k*. We prove it by contradiction. Assume that the maximum independent set of *G* is of size k' < k.

By Corollary 1, we can assume that  $t_1$  is the center sequence. By Lemma 1, we can further assume that the pair-wise alignment score between  $t_1$  and each of  $s_i$  for  $1 \le i \le n$  is exactly n + 3 and the alignment follows either Alignment (1) or (2). As the constrain  $b^k$  is satisfied, we assume the *x*th "b" in  $b^k$  appear in column  $\ell_x$  of the alignment matrix, in which the whole column should consist of "b" only. Suppose that column  $\ell_x$  intersects with the h(x)th "aba" of  $t_1$ , and h(x)th "bab"/"baa"/"aab" of  $s_i$  for  $1 \le i \le n$ . We define a subset of vertex  $\Phi$  that includes all  $v_{h(x)}$  for  $1 \le x \le k$ .

We prove that  $\Phi$  is an independent set of *G*. For any two vertices  $v_{h(x)}$  and  $v_{h(y)}$ in  $\Phi$  for x < y, we claim that edge  $e_i = (v_{h(x)}, v_{h(y)})$  does not exist. If not, there is a sequence  $s_i = (bab)^{h(x)-1}(baa)(bab)^{h(y)-h(x)-1}(aab)(bab)^{n-h(y)}$ . It can be verified that to follow Alignment (1) or (2), there is no way to have both the "b" in "baa" and the "b" in "aab" of  $s_i$  appear in column  $\ell_x$  and column  $\ell_y$  of the alignment matrix, respectively. Thus  $s_i$ , as well as  $e_i$ , does not exist. Therefore,  $\Phi$  is an independent set of *G* and it is of size k > k', which is a contradiction.

By Lemma 2, we can reduce the NP-hard problem of MIS to CCSA with binary alphabet, and hence we have the following theorem.

**Theorem 3** Constrained Center-Star Sequence Alignment (CCSA) with binary alphabet is NP-Hard.

#### 3.2 Inapproximability of CCSA

In this part, we show that unless P = NP, no polynomial-time algorithm can solve the CCSA problem of an arbitrary distance function within a constant approximation ratio r > 0. If there is such an algorithm, we show that the MIS problem can be solved in polynomial time. In particular, we will focus on the CCSA problems where the distance function  $\delta$  does not satisfy the triangle inequality.

We show a new reduction from MIS to CCSA which is similar to that in Sect. 3.1. The new transformation  $\Pi'(G, k) = (\Sigma, S, c, \delta)$  is defined as follows. The alphabet set consists of one more character than before, i.e.,  $\Sigma = \{a, b, c\}$ . The set of sequence  $S = \{t_1, \ldots, t_m, s_1, \ldots, s_m\}$ , where  $t_1 = \cdots = t_m = c(aba)^n c$  and for each edge  $e_i = (v_p, v_q)$  with  $p < q \ s_i = c(bab)^{p-1}(baa)(bab)^{q-p-1}(aab)(bab)^{n-q}c$ . The

	а	b	С	_
a	0	1	2	В
b	1	0	2	В
с	2	2	0	0
-	В	В	0	0

constrained sequence is still  $C = b^k$ . The new distance function  $\delta$  between any two characters in  $\Sigma \cup \{-\}$  is defined as follows.

where  $B = r \cdot m(n+3) + 1$ . Note that the distance function may not satisfy the triangle inequality.

Similar to Lemma 2, we can show the new transformation  $\Pi'$  yields a polynomialtime reduction from MIS to CCSA.

**Lemma 3** There is an independent set for G of size k if and only if there is a constrained center sequence alignment for the transformed instance  $\Pi'(G, k)$  of CCSA with score m(n + 3).

In fact, for the transformed instance  $\Pi'(G, k)$ , one can obtain the optimal alignment if the guaranteed score is at most  $r \cdot m(n + 3)$ .

**Lemma 4** If there is a constrained center sequence alignment for the transformed instance  $\Pi'(G, k)$  of CCSA with score at most  $r \cdot m(n + 3)$ , then this alignment has score exactly m(n + 3).

*Proof* Consider the case that when there is a center constrained sequence alignment of  $\Pi'(G, k)$  with score at most  $r \cdot m(n + 3)$ . Since the distance between space character and *a* or *b* is greater than  $r \cdot m(n + 3)$ , the alignment between  $s_i$  and  $t_1$  (ignoring the "*c*") for any *i* must follow Alignment (1) or (2). Hence, the score of each pair-wise alignment is exactly n + 3, and thus the total alignment score is m(n + 3).  $\Box$ 

As a result, if there is an algorithm that gives a solution of  $\Pi'(G, k)$  with approximation ratio r, then we can determine if an independent set for G of size k exists or not. If the algorithm gives an alignment of score at most  $r \cdot m(n + 3)$ , then by Lemma 4 the optimal alignment has score m(n + 3) and then by Lemma 3, there is an independent set for G of size k. If the algorithm gives an alignment of score greater than  $r \cdot m(n + 3)$ , then the optimal alignment has score greater than m(n + 3) and then by Lemma 3, there is no independent set for G of size k. Therefore, we have the following theorem for the inapproximability of CCSA.

**Theorem 4** Unless P = NP, there is no polynomial-time algorithm for CCSA with constant approximation ratio.

# **4** Conclusion

We have studied the CPSA problem when the constrained sequence is a string of k identical characters and gave an  $O(n^2)$  algorithm when k = t, where t is the minimum

number of occurrences of that character in these two sequences, i.e, the largest number of that character in the constrained sequence. However, for some k < t and problem instances, this CPSA problem might take  $O(n^3)$  time. It is not sure whether this CPSA problem can be solved in strictly less than  $O(n^3)$  time for all k.

Two negative results of the CCSA problem were shown in this paper. However, whether there exist constant approximation algorithms for binary alphabets or for 3-letter alphabets with distance function satisfying triangle inequality are both interesting open problems.

Since the solution for the center-star alignment problem can provide a good approximation for the MSA problem, the solution for the CCSA problem can also give a good approximation for the CMSA problem. Unfortunately, the CCSA problem has been proved to be NP-complete and also difficult to find an approximate solution for some distance function. Thus, it remains open whether there exists a good approximation algorithm for the CMSA problem.

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