# String Factorization via Prefix Free Families 

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#### Abstract

A factorization of a string $S$ is a partition of $w$ into substrings $u_{1}, \ldots, u_{k}$ such that $S=u_{1} u_{2} \cdots u_{k}$. Such a partition is called equality-free if no two factors are equal: $u_{i} \neq u_{j}, \forall i, j$ with $i \neq j$. The maximum equality-free factorization problem is to find for a given string $S$, the largest integer $k$ for which $S$ admits an equality-free factorization with $k$ factors.

Equality-free factorizations have lately received attention because of their applications in DNA self-assembly. The best approximation algorithm known for the problem is the natural greedy algorithm, that chooses iteratively from left to right the shortest factor that does not appear before. This algorithm has a $\sqrt{n}$ approximation ratio (SOFSEM 2020) and it is an open problem whether there is a better solution.

Our main result is to show that the natural greedy algorithm is a $\Theta\left(n^{1 / 4}\right)$ approximation algorithm for the maximum equality-free factorization problem. Thus, we disprove one of the conjectures of Mincu and Popa (SOFSEM 2020) according to which the greedy algorithm is a $\Theta(\sqrt{n})$ approximation.

The most challenging part of the proof is to show that the greedy algorithm is an $O\left(n^{1 / 4}\right)$ approximation. We obtain this algorithm via prefix free factor families, i.e. a set of non-overlapping factors of the string which are pairwise non-prefixes of each other. In the paper we show the relation between prefix free factor families and the maximum equality-free factorization. Moreover, as a byproduct we present another approximation algorithm that achieves an approximation ratio of $O\left(n^{1 / 4}\right)$ that we believe is of independent interest and may lead to improved algorithms. We then show that the natural greedy algorithm has an approximation ratio that is $\Omega\left(n^{1 / 4}\right)$ via a clever analysis which shows that the greedy algorithm is $\Theta\left(n^{1 / 4}\right)$ for the maximum equality-free factorization problem.


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## 1 Introduction

A factorization of a string $S$ is a partition of $S$ into substrings $u_{1}, u_{2}, \ldots, u_{k}$ such that $S=u_{1} u_{2} \cdots u_{k}$. Factorizations are central objects of study in stringology, a famous example being the Lempel-Ziv algorithm [14]. String factorizations have many other applications as we show next. For instance, finding an occurrence of a string $v$ in a text $T$ can be formulated as $T$ admitting a factorization $T=u v w$. Then, a string $v$ is a prefix of another string $T$ if $T=v w$ and it is a suffix of $T$ if $T=u v$. Moreover, many string problems can be seen as string factorization problems [9] such as: Shortest Common Superstring, Longest Common Subsequence and Shortest Common Supersequence, to name a few. Another example of a string factorization problem is the Minimum Common String Partition [6, 7], a problem concerned with identifying factorizations for two strings such that the sequence of factors for one string is the permutation of the other's.

In this paper we focus on the equality-free factorization, a special case of string factorization in which all factors are distinct. The equality-free factorization is a restricted variant of a more famous problem, termed generalized function matching which has a long history starting from 1979 (see, e.g., [12] and the references therein for more details). In the generalized function matching the input consists of a text $t$ over an alphabet $\Sigma_{1}$ and a pattern $p=p_{1} p_{2} \ldots p_{m}$ over an alphabet $\Sigma_{2}$. The goal is to find an injective function from $f: \Sigma_{2} \rightarrow \Sigma_{1}^{+}$such that $t=f\left(p_{1}\right) f\left(p_{2}\right) \ldots f\left(p_{m}\right)$. Thus, the maximum equality free factorization problem is a particular case of the generalized function matching in which all the characters of the pattern $p$ are distinct. In turn, generalized function matching is a particular case of string equations, which is a notoriously difficult problem (see, e.g., the JACM paper [13]). In fact, even the version which restricts character-to-character function matching is extremely difficult, see [1] as opposed to the more restricted parameterized matching $[2,3,10]$ which is simpler. Thus, maximum equality free factorization is part of family of fundamental problems in stringology.

The maximum equality free factorization problem is also motivated by applications in DNA synthesis [4]. More specifically, it is possible to produce short DNA fragments that will self-assemble into the wanted DNA structure. However, to obtain the desired structure, it is required that no two fragments are identical. Since the fragments must be short, one approach is to split the target DNA sequence into as many distinct pieces as possible.

## Previous work

The equality-free factorization problem was first introduced by Condon, Maňuch and Thachuk [4] where it was presented as the string partitioning problem. The string partitioning problem asks for a factorization into distinct factors such that each factor is at most of a certain length. The problem was studied in a more general setting where the measure of collision between two factors is either equality or one is a prefix/suffix of the other. Condon et al. showed that these variants are $\mathcal{N} \mathcal{P}$-complete. More recently, Fernau, Manea, Mercass and Schmid [5] presented a similar problem that imposes a lower bound on the number of factors instead of an upper bound on factor length. Fernau et al. showed that this variant is also $\mathcal{N} \mathcal{P}$-complete. Afterwards, Schmid [9] studied the Fixed-Parameter Tractability of the two problems.

The decision version of the problem, that is, given a string $S$ and an integer $k$, decide if there exists an equality free factorization of $S$ with at least $k$ factors, is termed MAXEFF-s (this is the notation of Schmid [9] and we decide to use it for the sake of consistency). The optimization version, in which we are given $S$ and the goal is to find an equality free
factorization with as many factors as possible, is termed OPTEFF-s. The acronyms for the two problems were introduced in the previous papers [5, 9, 11] and we will use them in our paper for consistency (OPT stands for optimization).

Mincu and Popa [11] study OPTEFF-s and another variant named Maximum Gapped Equality Free Factorization (OPTGEFF-s). In the latter it is not required that all the characters of the input strings are part of the factorization. That is, the goal is to find an equality free factorization using a maximum number of factors of a substring of the input string. More formally, a gapped factorization of string $S$ over alphabet $\Sigma$ is a tuple $\left(u_{1}, u_{2}, \ldots, u_{k}\right)$ such that $S=\alpha_{0} u_{1} \alpha_{1} u_{2} \alpha_{2} \cdots \alpha_{k-1} u_{k} \alpha_{k}$, where $u_{i} \in \Sigma^{+}$are the factors and $\alpha_{i} \in \Sigma^{*}$ are the gaps. OPTGEFF-s asks, for a given string $S$, to find the largest integer $k$ such that $S$ admits a gapped equality-free factorization of size $k$. In [11] a 2-approximation for the OPTGEFF-s and a $\sqrt{n}$-approximation (where $n$ is the size of the input string) for the OPTEFF-s were shown. Moreover, it was conjectured [11] that the greedy approximation ratio is $\Omega(\sqrt{n})$. Grüttemeier et al. [8] show a randomized algorithm for solving the MAXEFF-s with running time $2^{k} \cdot k^{O(1)}+O(n)$.

## Our results

As mentioned, the best-known approximation algorithm, the greedy algorithm, for the OPTEFF-s has an approximation ratio of $\sqrt{n}$ and it was conjectured that the greedy algorithm has an approximation ratio of $\Theta(\sqrt{n})$. In this paper, we show a better approximation algorithm for OPTEFF-s with ratio $O\left(n^{1 / 4}\right)$. We then use this algorithm to show that the greedy algorithm has the same approximation ratio of $O\left(n^{1 / 4}\right)$. Hence, this disproves the conjecture from [11] saying that the approximation ratio of the greedy algorithm is $\Omega(\sqrt{n})$.

The challenge is to show that the greedy algorithm has an approximation ratio of $O\left(n^{1 / 4}\right)$. To get our approximation ratio we start with an (approximate) prefix free solution for the version with gaps. Then, we use the prefix free property to map the factors of a solution returned by the greedy algorithm to the aforementioned prefix free solution. Moreover, besides the greedy algorithm, we introduce another approximation algorithm for OPTEFF-s with an approximation ratio of $O\left(n^{1 / 4}\right)$, that uses some interesting techniques and is of independent interest. We claim that our techniques give some key insights and perhaps open the path for better approximation algorithms for the problem.

Finally, we use a clever analysis to also show that the greedy algorithm cannot have an approximation ratio better than $O\left(n^{1 / 4}\right)$ and, hence, the approximation ratio of the greedy algorithm is $\Theta\left(n^{1 / 4}\right)$ for the maximum equality-free factorization problem.

## 2 The prefix-free property

For the OPTGEFF-s problem (the version of factorization with gaps), a 2-approximation algorithm via a reduction from a scheduling problem was shown in [11]. A natural direction for proving an approximation algorithm for OPTEFF-s is to transform a factorization with gaps, obtained from the approximation algorithm of OPTGEFF-s, into a solution without gaps. However, it is difficult to transform a given factorization with gaps into a factorization without gaps with roughly the same number of factors.

In this section we show that it is possible to transform a special case of a factorization with gaps into a factorization without gaps. We introduce the notion of a prefix-free gapped factorization, which has an important role in our algorithm and might be of independent interest.

- Definition 1. Let $n, k \in \mathbb{N}$, let $S$ be a string of length $n$ and let $F_{k}=\left\{S_{1}, S_{2}, \ldots, S_{k}\right\}$ be a set of non-overlapping factors of $S$ (possibly with gaps). F is a prefix-free gapped factorization of $S$ if for all $i \neq j, S_{i}$ is not a prefix of $S_{j}$.

Given a prefix-free gapped factorization $F$ such that $|F|=k$, we prove that there is a transformation of $F$ into a factorization without gaps with the same number of factors $k$, since each factor $S_{i}$ can be extended until the next factor $S_{i+1}$ without colliding with another factor $S_{j}$.

- Lemma 2. Let $n, k \in \mathbb{N}$ and let $S$ be a string of length $n$ with a prefix-free gapped factorization $F$ with $|F|=k$. Then, there is an equality free factorization for $S$ without gaps with at least $k$ factors.

Proof. Denote $S=T_{0} S_{1} T_{1} S_{2} T_{2} S_{2} \ldots T_{k-1} S_{k} T_{k}$. Denote with $R_{i}=S_{i} T_{i}$. Note that for each $i \neq j, R_{i}$ and $R_{j}$ are not prefixes of each other because their prefixes are $S_{i}$ and $S_{j}$, respectively, which are not prefixes of each other.

Now, consider $S=T_{0} R_{1} R_{2} \ldots R_{k}$. If, for all $i, T_{0} \neq R_{i}$, then we have an equality free ( $k+1$ )-factorization. Otherwise, there exists an $i$ such that $T_{0}=R_{i}$. We distinguish two cases.

In the first case, if $i<k$, then we set $Q_{i}=R_{i} R_{i+1}$. Thus, $Q_{i}$ and all other factors $R_{j}$ are still not prefixes of each other. $T_{0}$, which equals $R_{i}$, is not a prefix of any other $R_{j}$ (because it equaled $R_{i}$ ) and is shorter than $Q_{i}$. Hence $S$ has a $k$-factorization $S=$ $R_{i} R_{1} R_{2} \ldots R_{i-1} Q_{i} R_{i+2} \ldots R_{k}$.

In the second case, if $i=k$, then we set $Q_{k}=R_{k-1} R_{k}$ and using a similar argument as above we obtain a $k$-factorization $S=R_{k} R_{1} R_{2} \ldots R_{k-2} Q_{k}$.

## 3 An $O\left(n^{1 / 4}\right)$-approximation algorithm

In this section we show an algorithm that has an $O\left(n^{1 / 4}\right)$-approximation to OPTEFF-s.
Our algorithm (see Algorithm 1) is composed of two algorithms: a greedy algorithm, called Greedy1 (see Algorithm 3), which always yields an $\sqrt{n}$-approximation, and a new algorithm (Algorithm 2) which is described next. Algorithm 1 simply selects the better of the two algorithms and returns it.

Algorithm 1 An $O\left(n^{1 / 4}\right)$-approximation algorithm for OPTEFF-s.

```
Input: String S.
    F}\leftarrow\mathrm{ Algorithm 2(S);
    G\leftarrowAlgorithm 3(S);
    if }|G|>|F| the
        return G
    return F
```

The basic idea behind Algorithm 2 is to find, for every fixed integer $1 \leq i \leq 2 \sqrt{n}$, a greedy equality free gapped factorization of the input string in which every factor has length exactly $i$. The algorithm chooses from these gapped fixed length factorizations, the factorization with the most factors. Then, due to Lemma 2, we append to each of these factors the following adjacent (possibly empty) gap and we obtain an equality free factorization. See Algorithm 2 for more details.

Algorithm 2 Fixed length greedy factorization.

```
Input: String \(S\).
\(F \leftarrow \emptyset ;\)
for \(i \leftarrow 1\) to \(2 \sqrt{n}\) do
        \(j \leftarrow 1, G \leftarrow \emptyset ;\)
        while \(j \leq n-i\) do
            if \(S[j . . j+i-1] \notin G\) then
                \(G \leftarrow G \cup\{S[j . . j+i-1]\} ;\)
                \(j \leftarrow j+i-1 ;\)
        \(j \leftarrow j+1 ;\)
        if \(|G|>|F|\) then
        \(F \leftarrow G ;\)
    Extend each factor \(w_{i} \in F\) until factor \(w_{i+1}\) (for the last factor, extend it until the
    end of \(S\) );
    return \(F\)
```

- Lemma 3. Algorithm 2 yields an equality-free factorization without gaps.

Proof. First, in the for loop, at each step, Algorithm 2 adds to $G$ only distinct substrings of S . Then, notice that for every two factors $w_{1}, w_{2} \in G$, it holds that $w_{1}$ is not a prefix of $w_{2}$, since both $w_{1}$ and $w_{2}$ have the same length. Therefore, $G$ is a prefix-free gapped factorization, and due to Lemma 2, the factors are extended as in line 11 to have an equality-free factorization without gaps.

## Analysis

Here we prove that when the optimal solution, denoted OPT, has "many" factors, Algorithm 2 returns a good approximation of OPT.

Formally, let $F$ be the factorization returned by Algorithm 2. Let $\alpha$ be $n /|O P T|$. Notice that $|O P T|=n / \alpha$. We claim that $|F|=\Omega\left(n / \alpha^{2}\right)$.

We first give an overview of the proof. First, we prove in Lemma 5 that there are at least $n / 2 \alpha$ short factors (of length at most $2 \alpha$ ) in OPT. Then we prove in Lemma 6 that there are at least $\Omega\left(n / \alpha^{2}\right)$ factors of the exact same length in OPT. Next, we prove in Lemma 9 that the factorization $F$ returned in Algorithm 2 is a 2-approximation of optimal fixed length factorization (see Definition 7). Finally, we prove in Lemma 10 that $|F|=\Omega\left(n / \alpha^{2}\right)$.

- Definition 4. An x-short factor of $S$ is a factor of length $\leq x$. An x-long factor of $S$ is a factor of length $>x$. When $x$ is clear we will simply call them short factors and long factors.
- Lemma 5. There exist at least $n / 2 \alpha$ factors in OPT that are $2 \alpha$-short.

Proof. Let LF denote the set of $2 \alpha$-long factors in OPT and SF denote set of the $2 \alpha$-short factors in OPT. We will use an argument on $n$, the length of $S$. Each long factor must be, by definition, of length $\geq 2 \alpha+1$. Hence, by length arguments, $|L F| \cdot(2 \alpha+1)+|S F| \cdot 1 \leq n$ and, hence, $|L F| \leq n /(2 \alpha+1)-|S F| /(2 \alpha+1)$. On the other hand, since $|O P T|=n / \alpha$, we have that $|S F|=n / \alpha-|L F|$. Putting these two equations together yields that $|S F|=n / \alpha-|L F| \geq$ $n / \alpha-n /(2 \alpha+1)+|S F| /(2 \alpha+1)$ and hence, $|S F|-|S F| /(2 \alpha+1) \geq n / \alpha-n /(2 \alpha+1)$ which in turn yields $2 \alpha|S F| /(2 \alpha+1) \geq(n \alpha+n) / \alpha(2 \alpha+1)$. Hence, $2 \alpha^{2}|S F| \geq n \alpha+n$ and $|S F| \geq n / 2 \alpha+n / 2 \alpha^{2} \geq n / 2 \alpha$.

Next, we show that among the short factors, $\Omega(1 / \alpha)$ fraction of them actually have exactly the same length.

- Lemma 6. There exists an integer $\ell \leq 2 \alpha$ such that there are at least $n / 4 \alpha^{2}$ short factors in OPT of length exactly $\ell$.

Proof. By Lemma 5, there are at least $n / 2 \alpha$ short factors in OPT. The average number of factors of each short length, is at least $\frac{n / 2 \alpha}{2 \alpha}=n / 4 \alpha^{2}$. By the pigeonhole principle, there exists an integer $\ell \leq 2 \alpha$ such that there are at least $n / 4 \alpha^{2}$ short factors in OPT of length exactly $\ell$.

Next we prove that Algorithm 2 is a constant approximation algorithm to the problem of gapped factorization with fixed lengths.

- Definition 7. The Fixed-Length Maximum Gapped Equality-Free Factorization Size (FLOptGEFF-s) problem is defined as follows. For a given string $S$ and an integer $r$, find the largest integer $m$, such that $S$ admits a gapped equality-free factorization of size $m$ where all factors are of length $r$.

In [11], the problem of OPTGEFF-s is reduced to the Job Interval Selection Problem with $k$ intervals (JISPk, see Theorem 8), which has a 2-approximation algorithm.

- Theorem 8 (restated from [11]). Given $n$ jobs containing $k$ time intervals each, find the maximum number of intervals that can be selected such that (i) no two intervals intersect and (ii) at most one time interval is selected per job.

Analogously to [11], FLOptGEFF-s is reducible as well to JISPk, and here we briefly show the reduction.

- Lemma 9. Algorithm 2 is a 2-approximation for FLOptGEFF-s.

Sketch Proof. We construct an instance of JISPk with $O(n)$ jobs from a string $S$ with $n$ characters. For each distinct substring of $S$ with length $r$, we create a job. For each substring $s$ we add $[a, b)$ as a time interval of $s$ for all occurrences $s=S[a, b]$ in $S$.

Since JISPk has a 2-approximation algorithm, we have that FLOptGEFF-s has a 2approximation algorithm as well. Moreover, the algorithm that approximates FLOptGEFF$\mathrm{s}(S, r)$ for some string $S$ and integer $r$ is in fact the inner loop of Algorithm 2, on the iteration where $i=r$.

We are ready to prove a lower bound on the number of factors returned by Algorithm 2.

- Lemma 10. Let $F$ be the factorization returned by Algorithm 2. Then, $|F|=\Omega\left(n / \alpha^{2}\right)$.

Proof. By Lemma 6, there exists $\ell$ such that there are at least $n / 4 \alpha^{2}$ short factors in OPT of length $\ell$.

Let $S_{A L G}^{\ell}$ be the number of factors of length $\ell$ produced by Algorithm 2, let $S_{G E F F}^{\ell}$ be FLOptGEFF-s $(S, \ell)$, and let $S_{O P T}^{\ell}$ be the number of factors of length $\ell$ in $O P T$.

By Lemma 9, there is a polynomial algorithm that is a 2 -approximation of the number of occurrences of a factor of length $\ell$ in $O P T$. Moreover, the algorithm behind Lemma 9 is in fact the inner loop of Algorithm 2. Notice that $\ell \leq 2 \alpha \leq 2 n /|O P T| \leq 2 \sqrt{n}$, since $|O P T| \geq \sqrt{n}$ and therefore there is an iteration where $i=\ell$. Then,

$$
S_{O P T}^{\ell} / 2 \leq S_{G E F F}^{\ell} / 2 \leq S_{A L G}^{\ell}
$$

where the first inequality is due to the definition of FLOptGEFF-s and the second inequality is due to Lemma 9.

Hence, combining Lemma 6 and Lemma 9, for the iteration where $i=\ell$ on line 9, $|G|=S_{A L G}^{\ell} \geq\left(n / 4 \alpha^{2}\right) / 2$. Since the number of factors returned by the algorithm is at least $|G|$ (i.e. $\left.|F| \geq S_{A L G}^{\ell}\right)$, we have that $|F|=\Omega\left(n / \alpha^{2}\right)$.

As stated before, Algorithm 1 is composed by two algorithms, Algorithm 2 and Algorithm 3. Algorithm 3 was introduced in [11] as Greedy1 algorithm. In a nutshell, consider starting "at the left" of the string and adding the next shortest substring (distinct from the already selected factors) to the incumbent factorization at each step of the algorithm. See [11] for details.

## Algorithm 3 Greedy1.

```
Input: String \(S\).
    \(j \leftarrow 1, F \leftarrow \emptyset ;\)
    for \(i \leftarrow 1\) to \(n\) do
        if \(S[j . . i] \notin F\) then
            \(F \leftarrow F \cup S[j . . i] ;\)
            \(j \leftarrow i+1 ;\)
    Extend the last factor of \(F\) until the end of \(S\);
    return \(F\)
```

It was proven in [11] that Greedy1 yields an equality-free factorization. Moreover, they prove that Greedy 1 produces at least $\Omega(\sqrt{n})$ factors.

- Theorem 11. Algorithm 1 is an $O\left(n^{1 / 4}\right)$-approximation polynomial time algorithm for the OPTEFF-s problem.

Proof. Combining Greedy1 with Algorithm 2, we have an algorithm that produces at least $\Omega\left(\max \left(\left(n / \alpha^{2}\right), \sqrt{n}\right)\right)$ factors. This gives an approximation ratio of $O\left(\min \left(\frac{n / \alpha}{n / \alpha^{2}},(n / \alpha) / \sqrt{n}\right)\right)=O(\min (\alpha, \sqrt{n} / \alpha))$, which is maximized at $\alpha=\sqrt{n} / \alpha$, i.e. at $\alpha=n^{1 / 4}$ 。

Finally, notice that the both Greedy1 and Algorithm 2 run in polynomial time of at most $O\left(n^{1.5} \log n\right)$.

## 4 The natural greedy algorithm is a $\Theta\left(n^{1 / 4}\right)$-approximation

In this section we prove that Greedy1 itself achieves an approximation ratio of $O\left(n^{1 / 4}\right)$.

- Lemma 12. Greedy1 is a 2-approximation of Algorithm 2.

Proof. Let $S$ be a string, and let $\ell$ be a positive integer. Let $F_{\ell}$ be a fixed $\ell$ length gapped factorization on $S$ such that $\left|F_{\ell}\right|=$ FLOptGEFF-s $(S, \ell)$. Let $F_{G}$ be the factorization that is the output of the Greedy1 algorithm. We show that $\left|F_{G}\right| \geq\left|F_{\ell}\right| / 2$.

We map each factor of $F_{\ell}$ to a factor of $F_{G}$ as follows. Let $f \in F_{\ell}$ be a factor in $F_{\ell}$ and let $i \leq j$ be two indices such that $f=S[i, j]$. In $F_{G}$, denote the factors that cover $S[i]$ and $S[j]$ as $g_{i}$ and $g_{j}$, respectively. If $g_{i} \neq g_{j}$, map $f$ to $g_{j}$. If $f$ is a suffix of $g_{j}$, then also map $f$ to $g_{j}$. Otherwise, $f$ is fully contained in a factor of $F_{G}$ and $f$ is not a suffix of $g_{j}$. Then, it must be the case that there is a factor $g_{s}$ in $F_{G}$ such that the suffix of $g_{s}$ is exactly $f$, as otherwise Greedy1 would have cut the factor $g_{j}$ right after index $j$, but $f$ is not a suffix of $g_{j}$. Therefore, map $f$ to $g_{s}$ (if there are more than one factors with $f$ as a suffix in $F_{G}$, map to one of them arbitrarily).

Now, let $g$ be a factor in $F_{G}$, and let $\hat{i} \leq \hat{j}$ be two indices such that $g=S[\hat{i}, \hat{j}]$. We claim that there are at most two factors in $F_{\ell}$ that are mapped to $g$. There is at most one factor in $F_{\ell}$ that overlaps $S[\hat{i}]$, and therefore mapped to $g$ because of overlapping two factors in $F_{G}$. Moreover, since all the factors in $F_{\ell}$ have the same size, there is at most one factor in $F_{\ell}$ such that the suffix of $g$ is equal to the factor. Therefore, there are at most 2 factors in $F_{\ell}$ that are mapped to $g$. To conclude, there are at most $2 \cdot\left|F_{G}\right|$ factors in $F_{\ell}$, for every $\ell$.

Let $F_{A}$ be the factorization returned by Algorithm 2. There is an $\ell$ such that $\left|F_{\ell}\right| \geq\left|F_{A}\right|$. Since we proved that $\left|F_{G}\right| \geq\left|F_{\ell}\right| / 2$ for every $\ell$, we also have that $\left|F_{G}\right| \geq\left|F_{A}\right| / 2$.

Combining the above lemma with the Theorem 11, we conclude the following theorem.

- Theorem 13. The approximation ratio of Greedy1 is $O\left(n^{1 / 4}\right)$.

Proof. By Lemma 12, Greedy1 is a constant approximation of Algorithm 2 and therefore Greedy 1 is also a constant approximation of Algorithm 1 (that simply uses Algorithm 2 and Greedy1 and returns the maximum between them). Since by Theorem 11 Algorithm 1 is an $O\left(n^{1 / 4}\right)$-approximation for OPTEFF-s we have that Greedy1 is also an $O\left(n^{1 / 4}\right)$-approximation.

## 5 Tightness of Algorithm 1

In this section we prove that our analysis of Algorithm 1 is actually tight. We show that there is a case where both Greedy1 and Algorithm 2 have an approximation ratio that is at least $\Omega\left(n^{1 / 4}\right)$.

Similar to [11], we define a string $S$ as follows. Let $n$ be a square of an even number, i.e. there is an integer $k$ such that $n=(2 k)^{2}$. Let $\Sigma=\left\{x_{1}, x_{2}, \ldots, x_{\sqrt{n}}\right\}$ be an alphabet. We define variables $X_{1}, X_{2}, \ldots, X_{\sqrt{n}}$ such that for each variable $X_{i}$, define $X_{i}=x_{1} x_{2} \ldots x_{i}$. The string $S$ is defined as $S=X_{1} X_{2} . . X_{\sqrt{n}}$. Note that $|S|=\Theta(n)$.

- Lemma 14. There exists a factorization of $S$ with $\Omega\left(n^{3 / 4}\right)$ factors.

Proof. We first factorize $S$ into $\Omega\left(n^{3 / 4}\right)$ factors with gaps, and afterwards we get rid of the gaps. We factorize the variables $X_{1} . . X_{\sqrt{n} / 2-1}$ using only one factor. Then, the variable $X_{\sqrt{n} / 2}$ is factorized into $x_{1} ; x_{2} ; . . ; x_{\sqrt{n} / 2}$, for a total of $\sqrt{n} / 2$ factors. At least $\sqrt{n} / 2-1$ factors are produced by the variables $X_{\sqrt{n} / 2+1} X_{\sqrt{n} / 2+2}$ as follows. The variable $X_{\sqrt{n} / 2+1}$ is factorized into $x_{1} x_{2} ; x_{3} x_{4} ; \ldots$, for a total of $\left\lfloor\left|X_{\sqrt{n} / 2+1}\right| / 2\right\rfloor$ factors. The variable $X_{\sqrt{n} / 2+2}$ is factorized into $x_{2} x_{3} ; x_{4} x_{5} ; \ldots$, also for a total of at least $\left\lfloor\left(\left|X_{\sqrt{n} / 2+1}\right|\right) / 2\right\rfloor$ factors.

In general, at each iteration $i$, the algorithm produces factors of length $i$ using $i$ variables and $i$ offsets. Each variable is of length at least $\sqrt{n} / 2$, therefore at least $i \cdot\lfloor(\sqrt{n} / 2) / i\rfloor \geq$ $\sqrt{n} / 2-i$ factors are produced by $i$ variables. For each iteration $i$, the $r$ th variable $X_{j}$ of iteration $i$ produces factors of length $i$ starting at index $r$ with respect to the beginning of $X_{j}$. This procedure produces an equality free factorization with gaps.

There is a constant $c>0$ such that there are $c \sqrt{\sqrt{n} / 2}=c n^{1 / 4}$ iterations to the process. Therefore, at least

$$
\sum_{i=1}^{c n^{1 / 4}} \sqrt{n} / 2-i \geq c n^{3 / 4} / 2-c^{2} \sqrt{n}=\Omega\left(n^{3 / 4}\right)
$$

factors are produced in this process.

We are left with handling the gaps. Notice that there are two reasons for a gap to occur. First, (1) on iteration $i$ and variable $X_{j}$, we produce $\left\lfloor\left|X_{j}\right| / i\right\rfloor$ factors, and $\left|X_{j}\right|-i\left\lfloor\left|X_{j}\right| / i\right\rfloor>0$, so we have a gap at the end of $X_{j}$. Second, (2) on iteration $i$ and the $r$ th variable of the iteration $X_{j}$, when $X_{j}$ is not the first variable of iteration $i(r \neq 1)$, the factorization of $X_{j}$ does not start from $x_{1}$ but from $x_{r}$.

For gaps of type 1 , let $X_{j}$ be a variable that has a gap at the end of $X_{j}$. Then, if $j \neq \sqrt{n}$, we extend the first factor of $X_{j+1}$ backwards to close the gap. This extended factor is unique since there is only one instance of $x_{j} x_{1}$ in $S$. If $j=\sqrt{n}$ and we are in the last variable, we extend the last factor of $X_{j}$. This extended factor is unique since there is only one instance of this length in $S$.

For gaps of type 2 , let $X_{j}$ be a variable that has a gap at the beginning of $X_{j}$. Then, we extend the last factor of $X_{j-1}$ forward to close the gap. This extended factor is unique since there is only one instance of $x_{j-1} x_{1}$ in $S$.

For gaps with both types 1 and 2 , we just handle them as gaps with type 1 .
On string $S$, Greedy1 produces $\Theta(\sqrt{n})$ factors. Hence, and by Lemma 14, we have the following corollary.

- Corollary 15. The approximation ratio of Greedy1 is $\Omega\left(n^{1 / 4}\right)$.

On the string $S$ described above, Algorithm 2 produces $O(\sqrt{n})$ factors. To see this, let $l$ be some length that is being observed in line 2 of Algorithm 2. There are (at most) $\sqrt{n}$ $x_{1}$ 's in string $S$. Therefore there are at most $\sqrt{n}$ factors containing $x_{1}$. On the other hand, every factor that does not contain $x_{1}$, must start in a unique character (since before the extension, every factor is of length exactly $l$ ). There are (at most) $\sqrt{n}$ unique characters in $S$. Therefore, there are at most $\sqrt{n}$ factors not containing $x_{1}$.

Hence, and by Lemma 14, we have the following corollary.

- Corollary 16. The approximation ratio of Algorithm 2 is $\Omega\left(n^{1 / 4}\right)$.

Finally, since both lower bounds were based on the same case of string $S$, we have the following corollary.

- Corollary 17. The approximation ratio of Algorithm 1 is $\Omega\left(n^{1 / 4}\right)$.


## 6 Conclusions and future work

In this paper we disproved one of the conjectures of Mincu and Popa [11] and show that the natural greedy algorithm for the OPTEFF-s problem has a $\Theta\left(n^{1 / 4}\right)$-approximation factor. It is, of course, a natural open question to improve the approximation ratio for OPTEFF-s using a different algorithm than the greedy. We believe that the key in succeeding to obtain such an algorithm is finding a better approximation algorithm for the case when the number of factors in an optimal solution is relatively small. Thus, the ideas introduced in Section 3, where we present an alternative $O\left(n^{1 / 4}\right)$ approximation algorithm, represent a promising direction.

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