

Advanced Algorithmics

Strategies for Tackling Hard Problems

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Lecture 7

2017-05-11

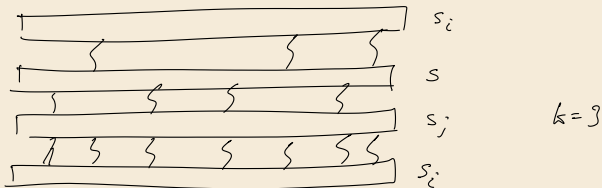
Depth-Bounded Search for Closest String

```
1 procedure closestStringFpt( $s, d$ ):
2   if  $d < 0$  then return "not found"
3   if  $d_H(s, s_i) > k + d$  for an  $i \in \{1, \dots, m\}$  then
4     return "not found"
5   if  $d_H(s, s_i) \leq k$  for all  $i = 1, \dots, m$  then return  $s$ 
6   Choose  $i \in \{1, \dots, m\}$  arbitrarily with  $d_H(s, s_i) > k$ 
7    $P := \{p : s[p] \neq s_i[p]\}$ 
8   Choose arbitrary  $P' \subseteq P$  with  $|P'| = k + 1$ 
9   for  $p$  in  $P'$  do
10     $s' := s$ 
11     $s'[p] := s_i[p]$ 
12     $s_{ret} := \text{closestStringFpt}(s', d - 1)$ 
13    if  $s_{ret} \neq \text{"not found"}$  then return  $s_{ret}$ 
14  return "not found"
```

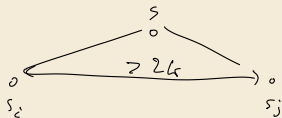
$\text{closestStringFpt}(s_1, k)$

Lemma 3.42 (Pair Too Different \rightarrow No)

Let $S = \{s_1, s_2, \dots, s_m\}$ a set of strings and $k \in \mathbb{N}$. If there are $i, j \in \{1, \dots, m\}$ with $d_H(s_i, s_j) > 2k$, then there is no string s with $\max_{1 \leq i \leq m} d_H(s, s_i) \leq k$.



d_H is a metric.



- $d_H(x, y) \geq 0$ ✓
- $d_H(x, x) = 0$ ✓
- Δ -ineq. $d_H(x, z) \leq d_H(x, y) + d_H(y, z)$



□

Theorem 3.43 (Search Tree for Closest String)

There is a search tree of size $O(k^k)$ for problem p -CLOSEST-STRING. & solves the problem. ◀

```
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```

Facts s can differ at at most k positions from any input.

If any candidate s has too many differences to any $s_i \rightarrow \downarrow$

W.l.o.g. $L \leq mk$ and all columns dirty.

Size of Search Trees

depth $\leq k$

fan-out $\leq k + 1$

\Rightarrow search space $\leq (k + 1)^k = O(k^k)$

$$\frac{(k+1)^k}{k^k} = \left(1 + \frac{1}{k}\right)^k < e$$

Correctness if we return a string ✓

if we abort : to show ~~A~~ consensus

Assume s_1 is not directly a consensus, but \exists consensus \hat{s}

$$\Rightarrow \forall i \quad d_H(s_1, s_i) \leq 2k$$

L. 3.42

$$s_i \text{ with } d_H(s_i, s_1) > k \quad P := \{j : s_1[j] \neq s_i[j]\}$$

$$|P| \leq 2k$$

$$|P'| = k+1 \quad P' \subseteq P$$

We call j correct if $j \in P_1 = \{p : s_1[p] \neq s_i[p] \wedge s_i[p] = \hat{s}[p]\}$

otherwise $j \in P_2 = \{p \in P : s_i[p] \neq \hat{s}[p]\}$

$$P = P_1 \dot{\cup} P_2 \quad d_H(\hat{s}, s_i) \leq k \rightsquigarrow |P_2| \leq k$$

\Rightarrow at least one j in P' is correct. ($P' \cap P_1 \neq \emptyset$)

\Rightarrow There is a path of modification to \hat{s} .

Corollary 3.44 (Closest String is FPT)

p -CLOSEST-STRING can be solved in time $\mathcal{O}(\underbrace{mL} + \underbrace{mk^2 \cdot \overbrace{k^k}^{\text{recursive calls}}})$.

↑

"preprocessing"

deleting non-dirty columns
and check > $m \cdot k$ columns remain

3.5 Interleaving

Up to now, considered two-phase algorithms

1. Reduction to problem kernel
2. Solve kernel by depth-bounded exhaustive search

Idea: Apply kernelization *in each recursive step*.

Setting for Interleaving

Assumptions: (more restrictive than general kernelization!)

- ▶ K kernelization that
 - ▶ produces kernel of size $\leq q(k)$ for q a polynomial (closest string art)
 - ▶ in time $\leq p(n)$ for p a polynomial
- ▶ Branch in depth-bounded search tree
 - ▶ into i subproblems with branching vector $\vec{d} = (d_1, \dots, d_i)$
(i.e. parameter in subproblems $k - d_1, \dots, k - d_i$)
 - ▶ Branching is computed in time $\leq \underline{r(n)}$ for r a polynomial
- ▶ search space has size $\mathcal{O}(\alpha^k)$.

\rightsquigarrow Running time of two-phase approach on input x with $n = |x|$ and $k = \kappa(x)$:

$$\mathcal{O}\left(\overbrace{p(n)}^{K(k)} + \underbrace{r(q(k))}_{\substack{\text{time per} \\ \text{recursive call}}} \cdot \underbrace{\alpha^k}_{\substack{\text{\# recursive calls}}}\right)$$

With Interleaving

Now replace splitting by:

c being parameter

-
- 1 **if** $|I| > c \cdot q(k)$ **then**
 - 2 $(I, k) := (I', k')$ where (I', k') forms a problem kernel // *Conditional Reduction*
 - 3 **end;**
 - 4 replace (I, k) with $(I_1, k - d_1), (I_2, k - d_2), \dots, (I_i, k - d_i)$. // *Branching*
-

\rightsquigarrow Running time of interleaved approach on input x with $n = |x|$ and $k = \kappa(x)$ is at most T_k :

$$\begin{array}{l} \text{Then } \exists \epsilon \text{ s.t.} \\ T_k \stackrel{\downarrow}{=} \mathcal{O}(\alpha^k) \end{array} \quad T_\ell = T_{\ell-d_1} + \dots + T_{\ell-d_i} + \underbrace{p(q(\ell)) + r(q(\ell))}_{\text{polynomial in } \ell}$$

Compare to non-interleaved version:

$$\begin{array}{l} T'_\ell = T'_{\ell-d_1} + \dots + T'_{\ell-d_i} \quad \rightsquigarrow \quad T'_\ell = \mathcal{O}(\alpha^\ell) \\ T_\ell = T_{\ell-d_1} + \dots + T_{\ell-d_i} + r(q(k)) \quad T_\ell = (r(q(k)) \cdot \alpha^k) \end{array}$$

Here the inhomogeneous term is constant w.r.t. ℓ , but depends on k

\rightsquigarrow cannot ignore constant factors

Theorem 3.45 (Linear Recurrences II)

Let $d_1, \dots, d_i \in \mathbb{N}$ and $d = \max d_j$.

Consider the inhomogeneous linear recurrence equation

$$T_n = T_{n-d_1} + T_{n-d_2} + \dots + T_{n-d_i} + \underline{\underline{f_n}}, \quad (n \geq d)$$

with $(f_n)_{n \in \mathbb{R}_{>0}}$ a known sequence of positive numbers and d initial values

$T_0, \dots, T_{d-1} \in \mathbb{R}_{>0}$.

$$f_n = \mathcal{O}(n^c)$$

Let $\underline{z_0}$ be the root with largest absolute value of $z^d - \sum_{j=1}^i z^{d-d_j}$ and assume $\underline{f_n = \mathcal{O}((z_0 - \varepsilon)^n)}$ for some fixed $\varepsilon > 0$.

Then $T_n = \mathcal{O}(T_n^0)$ where T_n^0 is defined as T_n with $f_n \equiv 0$. ◀

$$T(z) = \sum_{n \geq 0} T_n z^n = \underbrace{T_0 z^0 + \dots + T_{d-1} z^{d-1}}_{P(z)} + \sum_{j=1}^i \sum_{n \geq d_j} z^n T_{n-d_j} + \underbrace{\sum_{n \geq d} z^n f_n}_{F(z)}$$

$$= P(z) + F(z) + \left(\sum_{j=1}^i z^{d_j} \right) T(z) - Q(z)$$

$$T(z) = \frac{P(z) + Q(z) + F(z)}{1 - \sum_{j=1}^i z^{d_j}} = N(z)$$

$$= D(z)$$

$$N(z) = \text{polynomial} + F(z)$$

$F(z)$ is analytic $|z| < z_0$

$$\implies T_n = N(z_0) \cdot C / (\mu-1)! \cdot z_0^{-n} n^{\mu-1} (1 \pm \mathcal{O}(n^{-2}))$$

A Little Excursion: Singularity Analysis

$$f(z) = \frac{N(z)}{D(z)}$$

$D(z)$ polynomial

z_0 real root of $D(z)$ with smallest absolute value

and $N(z)$ analytic for all $|z| < z_0$

↑
"smooth"

$$= N(z) \left(\text{partial fractions of } \frac{1}{D(z)} \right)$$

$$\sum_{z_\lambda} \sum_{j=1}^{\mu_\lambda} \frac{c_{\lambda,j}}{(1 - \frac{z}{z_\lambda})^j}$$

z_0 is singularity of $f(z)$
(point where complex derivative)

approximate $f(z)$

for z near z_0

$$= \left[N(z_0) \pm O\left(1 - \frac{z}{z_0}\right) \right] \cdot \left(\frac{c_{0,\mu}}{\left(1 - \frac{z}{z_0}\right)^\mu} \pm O\left(\left(1 - \frac{z}{z_0}\right)^{-\mu+1}\right) \right)$$

$$= N(z_0) c_{0,\mu} \frac{1}{\left(1 - \frac{z}{z_0}\right)^\mu} \pm O\left(\left(1 - \frac{z}{z_0}\right)^{-\mu+1}\right) \quad (z \rightarrow z_0)$$

Taylor Expansion of f around z_0
 $f(z) = \sum_{i=0}^{\infty} \frac{f^{(i)}(z_0)}{i!} (z-z_0)^i$ analytic

$$= f(z_0) + f'(z_0)(z-z_0) \pm O((z-z_0)^2) \quad (z \rightarrow z_0)$$

$$\left(1 - \frac{z}{z_0}\right)^a = o\left(\left(1 - \frac{z}{z_0}\right)^b\right) \quad a > b$$

$$\left(\frac{\left(1 - \frac{z}{z_0}\right)^a}{\left(1 - \frac{z}{z_0}\right)^b} \right) = \left(1 - \frac{z}{z_0}\right)^{\overbrace{a-b}^{>0}} = 0 \quad z \rightarrow z_0$$

$$|z - z_0|^a = \left| \frac{z}{z_0} - 1 \right|^a z_0^a = z_0^a \left| 1 - \frac{z}{z_0} \right|^a$$

$$= \Theta \left(\left(1 - \frac{z}{z_0}\right)^a \right)$$

Theorem 3.46 (Transfer-Theorem of Singularity Analysis)

Assume $f(z)$ is Δ -analytic and admits the singular expansion

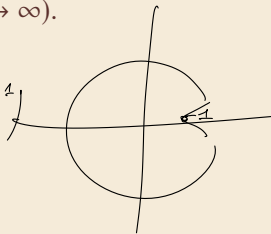
$$f(z) = g(z) \pm \mathcal{O}((1-z)^{-\alpha}) \quad (z \rightarrow 1) \quad \begin{array}{l} \nearrow z_0 \\ \leftarrow 1 \end{array}$$

with $\alpha \in \mathbb{R}$. Then

$$\begin{array}{l} f_n \text{ if } f(z) = \sum_n f_n z^n \\ // \\ [z^n]f(z) = [z^n]g(z) \pm \mathcal{O}(n^{\alpha-1}) \quad (n \rightarrow \infty). \end{array}$$

$$f(z) = \underbrace{N(z_0) \cdot c_{0,\mu}}_{\text{constant}} \frac{1}{\left(1 - \frac{z}{z_0}\right)^\mu} \pm \mathcal{O}\left(\left(1 - \frac{z}{z_0}\right)^{-\mu+1}\right)$$

$g(z)$



$$[z^n]f(z) = N(z_0) \cdot c_{0,\mu} \left([z^n] \frac{1}{\left(1 - \frac{z}{z_0}\right)^\mu} \right) \pm \mathcal{O}\left(z_0^{-n} n^{\mu-1}\right)$$

$$\begin{array}{l} // \\ \binom{n+\mu-1}{n} z_0^{-n} = \mathcal{O}\left(z_0^{-n} n^{\mu-1}\right) \end{array}$$

$$\begin{array}{l} f(z) = \sum_n f_n z^n \\ f(z_0) = \sum_n \underbrace{f_n z_0^n}_{f_n z_0^n} z_0^{-n} \end{array}$$

$$(1-z)^{-\mu} = \sum_{n \geq 0} \binom{-\mu}{n} (-z)^n (1)^{-\mu-n}$$

$$(a+b)^m = \sum_{n=0}^m \binom{m}{n} a^n b^{m-n}$$

$$\binom{-\mu}{n} = \frac{(-\mu)_n}{n!}$$

$$(z^{-1}) \leftarrow = \binom{-\mu}{n} (-1)^n$$

$$= \binom{n+\mu-1}{n}$$

$$\binom{n+\mu-1}{n} = \binom{n+\mu-1}{n-\mu+1} (-1)^n$$

$$\binom{r}{k} = \binom{k-r-1}{k} (-1)^k \quad k \in \mathbb{N}_0$$