## Window Subsequence Matching

# Window Subsequence Problems for Compressed Texts 

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## Outline of the Talk

- New topic in computer science: algorithms for compressed texts
- Our problem and our result
- Sketch of the algorithm

Straight-line Programs: Definition

## Example

Straight-line program (SLP) is a
Context-free grammar generating exactly one string Two types of productions:
$X_{i} \rightarrow a$ and $X_{i} \rightarrow X_{p} X_{q}$
abaababaabaab

$$
X_{1} \rightarrow b
$$

$X_{2} \rightarrow a$
$X_{3} \rightarrow X_{2} X_{1}$
$X_{4} \rightarrow X_{3} X_{2}$
$X_{5} \rightarrow X_{4} X_{3}$
$X_{6} \rightarrow X_{5} X_{4}$
$X_{7} \rightarrow X_{6} X_{5}$

INPUT:

Pattern: CES
Window size: 10
TASK: to find substrings of the length at most 10 in the text that contains CES as a subsequence

## OUTPUT:

C|O|M|P|U|T|E|R| $|\mathbf{S | C | I | I E | N | C | E | | | | N | ~}|$ R|U|S|S|||A
Problem for this talk:

How given a COMPRESSED text to solve window subsequence matching faster than just "unpack-and-search"?

## Part I

## What are compressed texts?

Can we do something interesting without unpacking?

SLP $=$ Compressed Text

Fact [Rytter, 2003]: given the archive of the text $T$ compressed by LZ78,LZW or some dictionary-based method of original length $n$ and the size of archive $z$ we can in time $O(z)$ convert it to SLP of size $O(z)$ generating the same text.

Fact [Rytter, 2003]: given the LZ77-compressed or RLE-compressed text $T$ of original length $n$ and the size of archive $z$ we can in time $O(z \log n)$ convert it to SLP of the size $O(z \log n)$ generating the same text.

Further by compressed text we mean an SLP generating it


## Why algorithms on compressed texts?

Answer for algorithms people:

- Might be faster than "unpack-and-search"
- Saving storing space and transmitting costs
- Many fields with highly compressible data: statistics (internet log files), automatically generated texts, massage sequence charts for parallel programs

Answer for complexity people:

- Some problems are hard in worst case. But they might be easy for compressible inputs
- New complexity relations. Similar problems becomes different

| $\exists$ poly algorithms: | At least NP-hard: |  |
| :--- | :--- | :--- |
| GKPR'96 Equivalence | L'06 Hamming distance | Part II |
| GKPR'96 Fully Compressed | LL'06 Fully Compressed |  |
| Pattern Matching | Subsequence Problem |  |
| GKPR'96 Regular Language | Lohrey'04 Context-Free | Our Problem and Our Result |
| Membership | Language Membership |  |
| GKPR'96 Shortest Period | LL'06 Longest Common Subsequence |  |
| L'06 Shortest Cover | BKLPR'06 Two-dimensional |  |
| L'06 Fingerprint Table | Compressed Pattern Matching |  |

## Window Subsequence Problems

Definition: $w$-window $=$ substring of the length $w$
Definition: minimal window $=$ substring containing the pattern, but any substring of which does not contain the pattern

INPUT: SLP generating text $T$, pattern $P$, window size $w$

## Computational tasks:

(1) To decide whether pattern $P$ is a subsequence of text $T$
(2) To compute the number of minimal windows of $T$ containing $P$
(3) To compute the number of $w$-windows of $T$ containing $P$

## Our Algorithm

## Main result:

Given a straight-line program of size $m$, a pattern of length $k$ and an integer $k$ we can solve all window subsequence problems on SLP-generated text in time $O\left(m k^{2} \log k\right)$

## Our Small Plan

- Define auxiliary data structures
- Compute them
- Derive answers for our tasks from these structures


## Window Subsequences: Motivation

Why do we do window subsequence matching (in compressed texts)?

- Variation of approximate pattern matching
- Useful for finding access patterns in databases
- Virus search in archives
- Pattern discovery in bioinformatics
- New step in the framework "what problems could be solved without unpacking?'


## Part III

## Algorithm for Window Problems on Compressed Texts

## Auxiliary Arrays

Let $X_{1}, \ldots, X_{m}$ be the nonterminals of SLP generating $T$, while $P_{1}, \ldots, P_{l}$ be all different substrings of pattern $P$

## Left inclusions

For every $X_{i}$ and every $P_{j}$ let us define $L(i, j)$ as the length of the minimal prefix of $X_{i}$ that contains $P_{j}$, in case of no such prefix exists let $L(i, j):=\infty$

Right inclusions
For every $X_{i}$ and every $P_{j}$ let us define $R(i, j)$ as the length of the minimal suffix of $X_{i}$ that contains $P_{j}$, in case of no such prefix exists let $R(i, j):=\infty$

## Auxiliary Arrays II

## Minimal windows

$M(i)=$ number of minimal windows containing $P$ in $X_{i}$

## Fixed windows

$F(i)=$ number of $w$-windows containing $P$ in $X_{i}$

## Computing Minimal Windows

We compute $M(i)$ by induction on $i$ and using already computed right/left inclusions:

Base: if $X_{i} \rightarrow a$, then $M(i)=0$ only except
$P=a$, in the latter case $M(i)=1$
Inductive step: $X_{i} \rightarrow X_{p} X_{q}$.
$M(i)=M(p)+M(q)+$ ???

## Computing boundary minimal windows

$\diamond$ Consequently consider decompositions $P=P_{u} P_{v}$
$\diamond$ For every decomposition with the help of $\mathrm{L} / \mathrm{R}$ inclusions info
$\diamond$ find the unique minimal window such that
$\diamond P_{u}$ is falling in $X_{p}$ and $P_{v}$ is falling $X_{q}$
$\diamond$ If this window is shifted, then we increment the counter
Complexity: $O(m k)$

## Summary

## Main points:

- Compressed text $=$ text generated by SLP
- Given SLP we can solve window subsequence matching in time $O\left(m k^{2} \log k\right)$
- Method: dynamic programming over SLP


## Open Problems:

- Decrease the $k$-depended factor in complexity
- To construct $O(n m)$ algorithms for edit distance, where $n$ is the length of $T_{1}$ and $m$ is the compressed size of $T_{2}$


## Computing Left Inclusions

We compute $L(i, j)$ by induction on $i$
Base: if $X_{i} \rightarrow a$, then $L(i, j)=\infty$ for all $P_{j} \neq a$,
and $L(i, j)=1$ in case $P_{j}=a$
Induction step: let $X_{i} \rightarrow X_{p} X_{q}$
If $L(p, j) \neq \infty$, then $L(i, j)=L(p, j)$. Assume $L(p, j)=\infty$.
If we find a decomposition $P_{j}=P_{u} P_{v}$ with minimal
$\left|P_{v}\right|$ where $L(p, u) \neq \infty$ and $L(q, v) \neq \infty$,
then we immediately get $L(i, j)=\left|X_{p}\right|+L(q, v)$
Such a decomposition can be found by a binary search
Total complexity $O\left(m k^{2} \log k\right)$
where $m$ is the size of SLP and $k$ is the length of the pattern
Mikhail Dvorkin: $O\left(m k^{2}\right)$

## Deriving the Answer

## Computational tasks:

- To decide whether $P$ is a subsequence of $T$

$$
\text { - Answer: "yes" iff } M(m) \neq 0
$$

- To compute the number of $w$-windows of $T$ containing $P$
- Answer: $F(m)$
- To compute the number of minimal windows of $T$ containing $P$
- Answer: $M(m)$

Complexity: $O\left(m k^{2} \log k\right)$.

## Last Slide

Yury Lifshits http://logic.pdmi.ras.ru/~yura/
Relevant papers:
R Yu. Lifshits
Solving Classical String Problems on Compressed Texts
Pu. Lifshits and M. Lohrey
Querying and Embedding Compressed Texts
P. Cégielski, I. Guessarian, Yu. Lifshits and Yu. Matiyasevich Window Subsequence Problems for Compressed Texts
L.Boasson, P. Cégielski, I. Guessarian, and Yu. Matiyasevic Window-Accumulated Subsequence Matching Problem is Linear

- P. Cégielski, I. Guessarian, and Yu. Matiyasevich

Multiple Serial Episode Matching
Thanks for attention!

