

Compression by Contracting Straight-Line Programs

Moses Ganardi



MAX PLANCK INSTITUTE
FOR SOFTWARE SYSTEMS

Algorithms on Compressed Data

Two goals in data compression:

- ◆ Store data in a **compact form** (lossless).
- ◆ Support **efficient queries** directly on the compressed representation.
→ Avoid decompression!

compression

algorithmics

Kolmogorov
complexity

LZ77

grammar-based
compression

RLE

uncompressed

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Grammar-based compression

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A **straight-line program (SLP)** is a context-free grammar \mathcal{G} which produces exactly one string.

Every variable occurs exactly once on the left-hand side of a rule and the variables are topologically ordered.

grammar

$$S_4 \rightarrow S_3 S_2$$

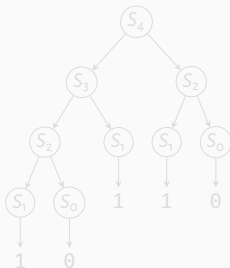
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$$S_1 \rightarrow 1$$

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derivation tree



The string length is denoted by $N \leq 2^{\Theta(|\mathcal{G}|)}$.

Chomsky normal form: rules of the form $A \rightarrow BC$ or $A \rightarrow a$.

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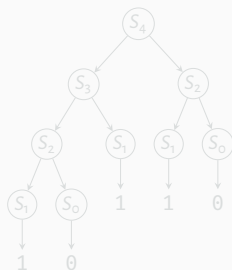
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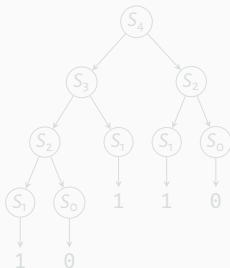
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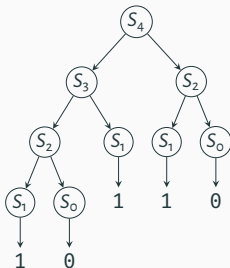
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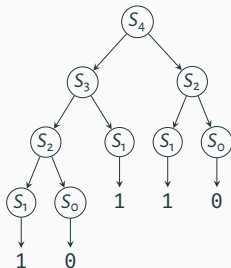
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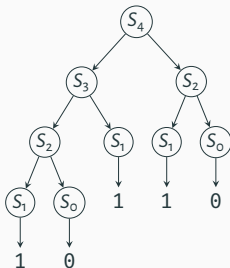
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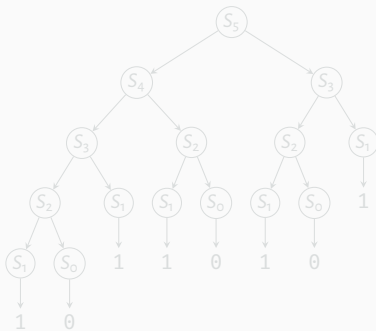
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The random access problem

For algorithmic applications the two important parameters are **size** and **height**.

Example: Random access in time $\mathcal{O}(\text{height})$:

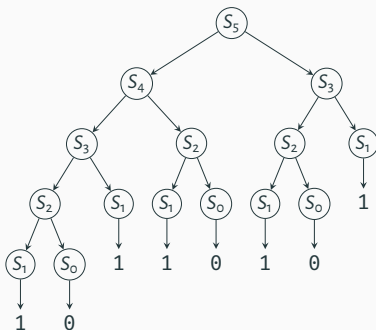


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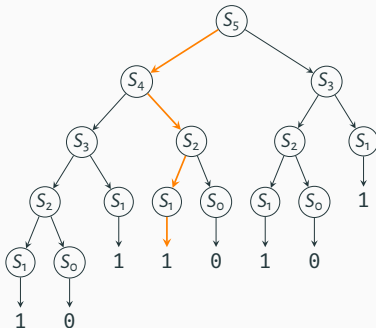


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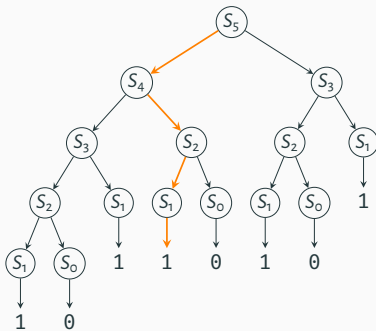


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Balancing Theorem (G, Jež, Lohrey, FOCS 2019, JACM 2021)

Given an SLP \mathcal{G} for a string of length N . One can compute in linear time an equivalent SLP of height $\mathcal{O}(\log N)$ and size $\mathcal{O}(|\mathcal{G}|)$.

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[Rytter, 2002; Charikar et al., 2002]

→ simple solution for random access in $\mathcal{O}(\log N)$ time and **linear** space

Other applications:

- ◆ rank and select queries, computing fingerprints, range minimum queries, subsequence matching
- ◆ spanner evaluation

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1. Can we refine the balancing theorem, establishing stronger balancedness properties “for free”? (= $\mathcal{O}(1)$ factor increase)
2. Which algorithmic applications can be obtained using such balancing results?

Compressed pattern matching

Does balancing lead to improved algorithms for **compressed pattern matching**?

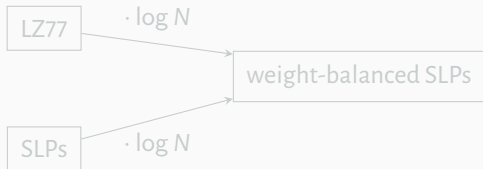
Given an uncompressed pattern P of length m , and
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Question Does P occur in T ?

Theorem (Gawrychowski, 2011)

Compressed pattern matching can be solved in time

- ◆ $\mathcal{O}(m + n \cdot \log N)$ for LZ77-compression and for SLPs,
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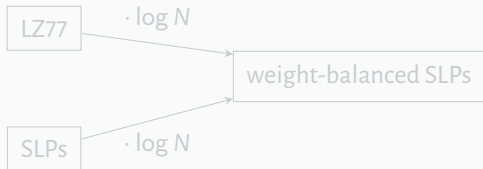
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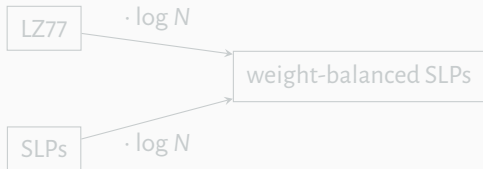
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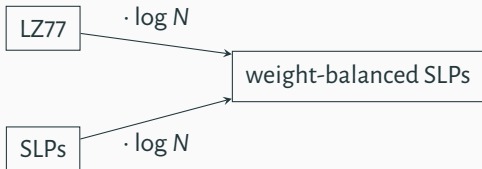
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Zoo of balanced SLPs



logarithmic height



weight balanced



height balanced

Question: Is the multiplicative cost of $\mathcal{O}(\log N)$ optimal?



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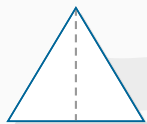
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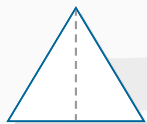
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Theorem

There exist SLPs of size $\mathcal{O}(n)$ such that any equivalent path balanced SLP has size $\Omega(n \log N)$.



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In every subtree T , every root-to-leaf path has length $\Theta(\log |T|)$.

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1. Compressed Pattern Matching

Theorem (G, Gawrychowski, 2022)

Compressed pattern matching for SLP-compressed texts can be solved in time $\mathcal{O}(m + n)$.

Relies only on logarithmic height SLPs (and new data structures).

2. A Refined Balancing Theorem

New algorithmic applications:

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Definition

An SLP is **contracting** if for every rule $A \rightarrow \beta_1 \dots \beta_k$ and every variable β_i we have $|\beta_i| \leq |A|/2$.

- ◆ Every variable A has height $\mathcal{O}(\log |A|)$ (**locally balanced**).
- ◆ Given a variable A , one can access $A[j]$ in time $\mathcal{O}(\log |A|)$.
- ◆ Useful when multiple strings s_1, \dots, s_m are compressed using a single SLP.

Theorem

One can convert an SLP \mathcal{G} in linear time into an equivalent contracting SLP of size $\mathcal{O}(|\mathcal{G}|)$ with rules of constant length.

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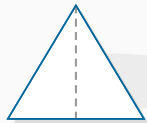
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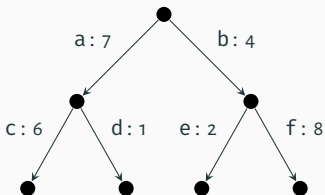


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Given a trie T with edges labeled by weighted symbols, define all prefixes by a **contracting** SLP.



Possible with a contracting SLP of size $\mathcal{O}(|T|)$.

Applications

In **finger search** on a (compressed) string we want to support the following operations:

[Bille, Christiansen, Cording, Gørtz, 2018]

- ◆ `setFinger(i)`
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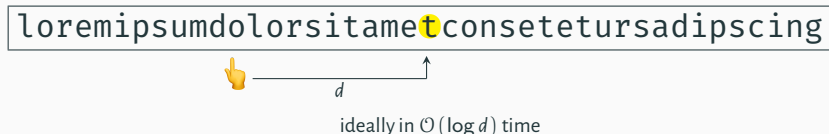


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ideally in $\mathcal{O}(\log d)$ time

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where d is the distance between i and the finger position, using $\mathcal{O}(|\mathcal{G}|)$ preprocessing time and space.

Choosing $t = \log^* N$ yields $\mathcal{O}(\log d)$ time for `access(i)` and `moveFinger(i)`.

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Theorem (Reh, Sieber, 2020)

Given a forest SLP for a tree T , one can support in linear space the following navigation steps on T in constant time:

- ◆ `parent()` in $\mathcal{O}(1)$ time
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Extend a navigation data structure on FSLP-compressed trees:

Theorem (Reh, Sieber, 2020; G, 2021)

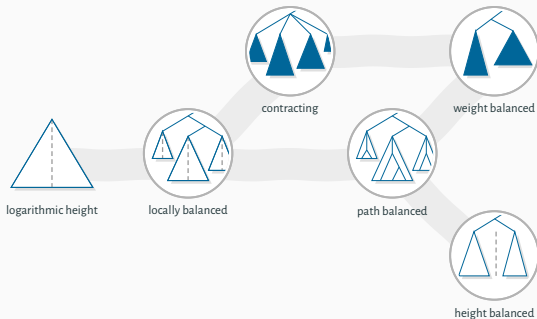
Given a forest SLP for a tree T , one can support in linear space the following navigation steps on T in constant time:

- ◆ `parent()` in $\mathcal{O}(1)$ time
- ◆ `first_child(), last_child()` in $\mathcal{O}(1)$ time
- ◆ `next_sibling(), prev_sibling()` in $\mathcal{O}(1)$ time
- ◆ `get_symbol()` in $\mathcal{O}(1)$ time
- ◆ `child(i)` in $\mathcal{O}(\log d)$ time

where d is the degree of the current node.

Conclusion

Balancing in grammar-based compression as a preprocessing step that enables fast queries on the compressed data.



Open questions:

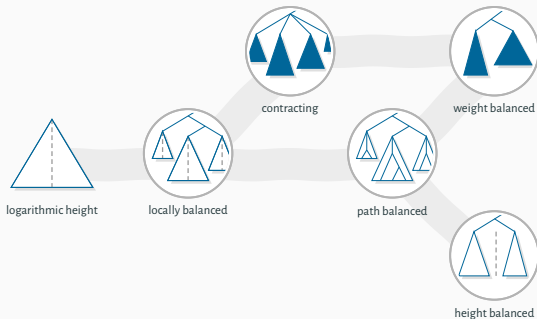
Finger search in $\mathcal{O}(\log d)$ time and $\mathcal{O}(|\mathcal{G}|)$ space?

random access for LZ77 in $\mathcal{O}(\log N)$ time and linear space?

Balancing for LZ77/collage systems?

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Finger search in $\mathcal{O}(\log d)$ time and $\mathcal{O}(|\mathcal{G}|)$ space?

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