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# Smallest grammar by recompression

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- related to LZW and LZ



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With  $\mathcal{O}(1)$  increase in size, this is an SLP.

## Definition (SLP: Straight Line Programme)

CFG with

- ordered nonterminals  $X_1, X_2, \dots$
- Chomsky normal form
- for  $X_i \rightarrow X_j X_k$  we have  $j, k < i$

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  - local replacement rules (plus a global partition): pairs and blocks
  - analysis vs LZ

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Linear time.



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Very simple linear-time algorithm,  $\mathcal{O}(\log(n/g))$  approximation.



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Algorithm similar to Sakamoto, different analysis.



# Example

*a a a b a b c a b a b b a b c b a*



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## Example

$a_3$  *b a b c a b a b b a b c b a*

$a_3 \rightarrow a^3$

## Example

$a_3$   $b$   $a$   $b$   $c$   $a$   $b$   $a$     $b_2$   $a$   $b$   $c$   $b$   $a$

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## Example

$a_3$   $b$   $d$   $c$   $d$   $a$   $b_2$   $d$   $c$   $b$   $a$

$a_3 \rightarrow a^3, b_2 \rightarrow b^2, d \rightarrow ab$

## Example

$a_3$   $b$   $d$   $c$   $d$   $a$   $b_2$   $d$   $c$   $e$

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### Intuition

- Phases: compress only pairs and block from the beginning of a phase.
- Treat nonterminals as letters.
- To speed up, we make some pair compression simultaneously (partition  $\Sigma$  to  $\Sigma_\ell, \Sigma_r$ , pairs from  $\Sigma_\ell \Sigma_r$ )

# Algorithm

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2:    $L \leftarrow$  list of letters in  $T$ 
3:   for each  $a \in L$  do                                ▷ Blocks compression
4:     compress maximal blocks of  $a$                        ▷  $\mathcal{O}(|T|)$ 
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8:   for  $ab \in P \cap \Sigma_\ell \Sigma_r$  do             ▷ These pairs do not overlap
9:     compress pair  $ab$                                    ▷ Pair compression
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10: return the constructed grammar
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# Partition

1/4 appearances covered

A partition  $\Sigma_\ell \Sigma_r$  such that 1/4 of pairs is covered.



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- After block compression  $aa$  does not appear.
- Random partition: 1/4 pairs can be covered.
- derandomise (expected value)
- we need number of appearances of  $ab$ : RadixSort
- $\mathcal{O}(|T|)$ .



# Size reduction

## Size drop

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## Towards running time

It is enough to show that one round runs in  $\mathcal{O}(|T|)$ .

# Running time

## Partition

$\mathcal{O}(|T|)$  time.

## Block compression

By RadixSort,  $\mathcal{O}(|T|)$  time.

## Pair compression

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  - first represent  $a^{\ell_2 - \ell_1}, a^{\ell_3 - \ell_2}, \dots, a^{\ell_k - \ell_{k-1}}$  as  $a_{\ell_2 - \ell_1}, a_{\ell_3 - \ell_2}, \dots, a_{\ell_k - \ell_{k-1}}$
  - do this by binary expansion  
(make new rules  $a_2 \rightarrow aa, a_4 \rightarrow a_2a_2, a_8 \rightarrow a_4a_4, \dots$ )

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(make new rules  $a_2 \rightarrow aa, a_4 \rightarrow a_2a_2, a_8 \rightarrow a_4a_4, \dots$ )
  - $a_{\ell_{i+1}} \rightarrow a_{\ell_{i+1} - \ell_i} a_{\ell_i}$
  - representation cost

$$\mathcal{O}\left(\sum_{i=1}^{k-1} \log(\ell_{i+1} - \ell_i)\right)$$

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- We begin with a  $G$  generating  $T$  (mental experiment)
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  - we only need to count the number of created credit



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## Definition (Non-crossing pairs)

$ab$  is **non-crossing pair** iff none of the below happens

- $aX$  appears in a rule,  $X$  begins with  $b$
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When each pair from  $\Sigma_\ell \Sigma_r$  is non-crossing, replace all those pairs in  $G$  (no new credit).

## Making pairs non-crossing

When  $ab$  has a crossing appearance:  $aX_i$  or  $X_ib$

- $X_i$  defines  $bw$ : change it to  $w$ , replace  $X_i$  by  $bX_i$
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### LeftPop( $b$ )

- 1: **for**  $i \leftarrow 1 \dots g - 1$  **do**
- 2:     **if** the first symbol in  $X_i \rightarrow \alpha$  is  $b$  **then**
- 3:         remove this  $b$
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### Lemma

*After LeftPop( $b$ ) and RightPop( $a$ ) the  $ab$  is non-crossing.*

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- Credit increases by  $\mathcal{O}(g)$



# Blocks & Wrap up

## Idea

Similarly as pairs

- $X_i$  defines  $a^{\ell_i} w b^{r_i}$ : change it to  $w$
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## In total

- $\mathcal{O}(g)$  per phase
- $\mathcal{O}(\log n)$  phases
- $\mathcal{O}(g \log n)$  credit in total (= size of created grammar)
- can be improved to  $\mathcal{O}(g \log(n/g))$

# Acknowledgments

M. Lohrey

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## P. Gawrychowski

- introducing to the topic
- literature
  - K. Mehlhorn, R. Sundar and Ch. Uhrig, *Maintaining Dynamic Sequences under Equality Tests in Polylogarithmic Time*, '97
  - H. Sakamoto, *A fully linear-time approximation algorithm for grammar-based compression*, '05
  - M. Lohrey and Ch. Mathissen, *Compressed Membership in Automata with Compressed Labels*, '11

# Open problems, related research

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- better approximation
- simpler computational model (no RadixSort)
- addition chains ( $\mathcal{O}(\frac{\log n}{\log \log n})$  approximation known)



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## Other applications: recompression

- compressed membership
- fully compressed pattern matching
- word equations

