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Smallest grammar by recompression

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17.06.2013

Grammar based-compression

Represent w as a CFG generating it.



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- it is usually small (at most quadratic vs. LZ)
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- related to LZW and LZ



Smallest grammar

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Given w return **smallest CFG** G_w such that $L(G_w) = w$.



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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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- translation of LZ into SLP, size $\mathcal{O}(\ell \log(n/\ell)) \leq \mathcal{O}(g \log(n/g))$
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 - local replacement rules (plus a global partition): pairs and blocks
 - analysis vs LZ



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Linear time.



This talk

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 a a b a b c a b a b b a b c b a



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Example

$a_3 \quad b \ a \ b \ c \ a \ b \ a \ b \ b \ a \ b \ c \ b \ a$
 $a_3 \rightarrow a^3$



Example

$a_3 \ b \ a \ b \ c \ a \ b \ a \ b_2 \ a \ b \ c \ b \ a$
 $a_3 \rightarrow a^3, b_2 \rightarrow b^2$



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 Σ to Σ_ℓ, Σ_r , pairs from $\Sigma_\ell \Sigma_r$)



Algorithm

```
1: while  $|T| > 1$  do
```



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1: while  $|T| > 1$  do  
2:    $L \leftarrow$  list of letters in  $T$   
3:   for each  $a \in L$  do  
4:     compress maximal blocks of  $a$ 
```

▷ Blocks compression
▷ $\mathcal{O}(|T|)$



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1: while  $|T| > 1$  do
2:    $L \leftarrow$  list of letters in  $T$ 
3:   for each  $a \in L$  do            $\triangleright$  Blocks compression
4:     compress maximal blocks of  $a$        $\triangleright \mathcal{O}(|T|)$ 
5:    $P \leftarrow$  list of pairs
6:   find partition of  $\Sigma$  into  $\Sigma_\ell$  and  $\Sigma_r$ 
7:    $\triangleright$  Try to maximize the occurrences from  $\Sigma_\ell \Sigma_r$  in  $T$ .

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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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8:   for  $ab \in P \cap \Sigma_\ell \Sigma_r$  do           ▷ These pairs do not overlap
9:     compress pair  $ab$            ▷ Pair compression
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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A partition $\Sigma_\ell \Sigma_r$ such that 1/4 of pairs is covered.

- 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

- Consider set of two consecutive letters ab in T .
- For 1/4 of them one letter is compressed in a phase.
- Length drops by a constant factor.



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When we consider ab we replace it, unless one letter was already replaced.
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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

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



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Representation cost



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- when $a^{\ell_1}, a^{\ell_2}, \dots, a^{\ell_k}$ are replaced with $a_{\ell_1}, a_{\ell_2}, \dots, a_{\ell_k}$ ($\ell_1 < \ell_2 \dots < \ell_k$):



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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 $a_2 \rightarrow aa, a_4 \rightarrow a_2a_2, a_8 \rightarrow a_4a_4, \dots$)
 - $a_{\ell_{i+1}} \rightarrow a_{\ell_{i+1} - \ell_i} a_{\ell_i}$



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 - $a_{\ell_{i+1}} \rightarrow a_{\ell_{i+1} - \ell_i} a_{\ell_i}$
 - representation cost

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



Analysis outline

- We begin with a G generating T (mental experiment)
- in each moment we keep G generating the current T



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- G is of more general form: $X_i \rightarrow uX_j vX_k w$
- explicit letters have credit
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 - we need 1 representation cost
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 - (bit more tricky for blocks)
- we only need to count the number of created credit



Pair compression

$X_1 \rightarrow ababcbab, X_2 \rightarrow abcbX_1 abX_1 a$



Pair compression

$X_1 \rightarrow abab\mathbf{cab}$, $X_2 \rightarrow abcbX_1\mathbf{ab}X_1a$

- compression of ab : easy



Pair compression

$X_1 \rightarrow ababca\textcolor{red}{b}$, $X_2 \rightarrow abcbX_1ab\textcolor{red}{X_1a}$

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

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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 $X_i b$

- 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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- Can be done in parallel for all $ab \in \Sigma_\ell \Sigma_r$.
- Credit increases by $\mathcal{O}(g)$



Blocks & Wrap up

Idea

Similarly as pairs

- X_i defines $a^{\ell_i}wb^{r_i}$: change it to w
- replace X_i in rules by $a^{\ell_i}X_ib^{r_i}$



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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

Suggesting the analysis.



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Acknowledgments

M. Lohrey

Suggesting the analysis.

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

