# An LMS-based Grammar Self-Index with Local Consistency Properties

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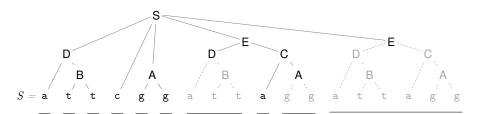
Let us denote the size of G as g and the number of nonterminals as r.

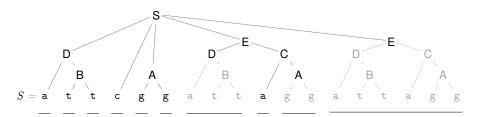
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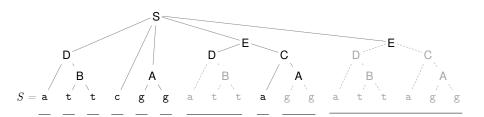
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Claude et al. 2020 demonstrated that there is a self-index on top of  $\mathcal G$  that requires  $O(g \lg n + (2+\epsilon)g \lg r)$  bits of space, where  $0<\epsilon\leq 1$  is a constant. This index can locate the occ the occurrences of a pattern P[1,m] in  $O((m^2+occ)\lg g)$  time.

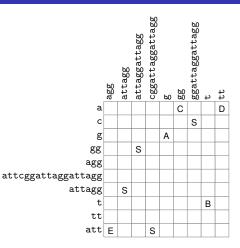


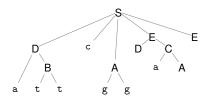


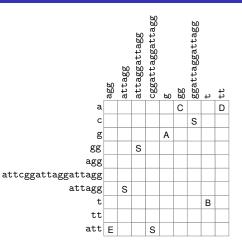


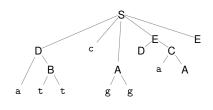
A→g g	gg	
C→a A	a gg	${\mathcal Y}$
S→D c AEE	att c gg attagg attagg	att
E→D C	att agg	cl
B→t t	t t	gg
D→a B	a tt	attagg

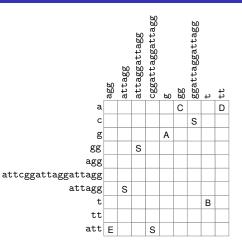
y X
att|cggattaggattagg
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 gg|attaggattagg
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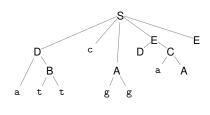




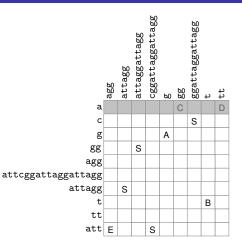


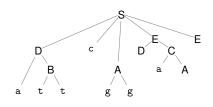




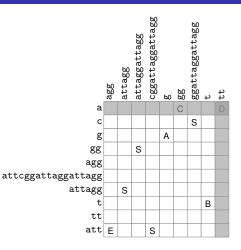


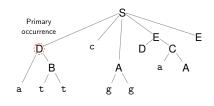
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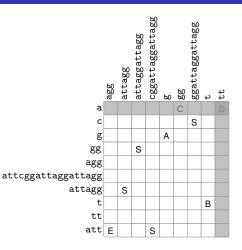


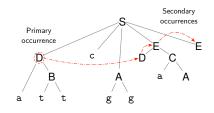
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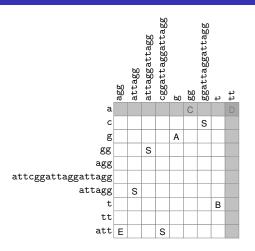


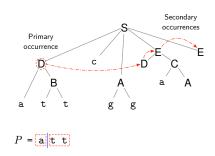
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**Problem**: We have to try out all the possible cuts of *P* 

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- It is balanced.
- The occurrences of a pattern P are largely compressed in the same way.

The algorithm of Christiansen et al. 2020 has several rounds of parsing. In the first round i = 1, we set  $S_i = S$  and apply the following procedure:

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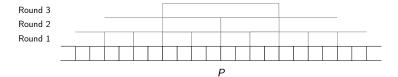
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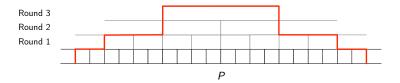
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- **5** Repeat the same parsing algorithm with  $S_{i+1}$

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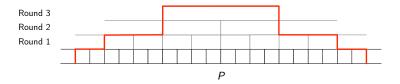


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The phrases below the red line are always the same, regardless of the context of P in S.

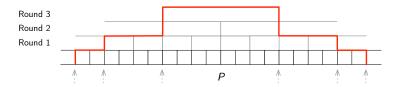
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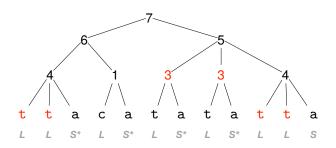
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- The resulting grammar can be potentially large compared to other heuristics, like RePair.
- It increases the size of the grammar index considerably.

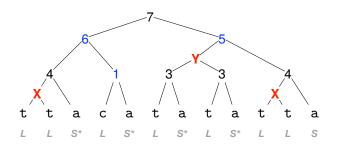
#### Our method

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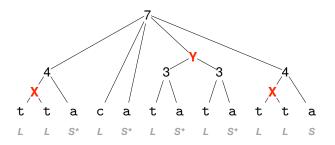
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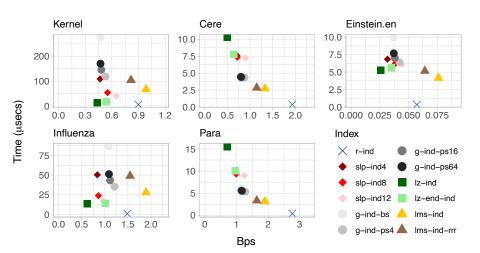
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  - r-ind: the r-index.
- We assessed the space usage and the time for answering the locate operation.

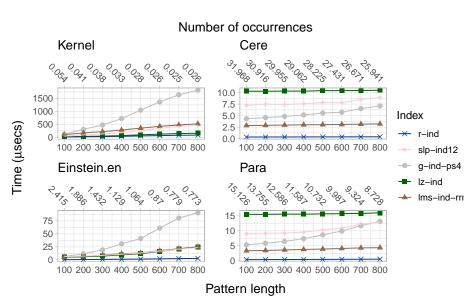
## Results: grammar algorithm

Dataset	n	$\sigma$	RePair	LMS	LMS post	LC
para	429,265,758	5	5,344,480	22,787,047	8,933,303	8,888,002
cere	461,286,644	5	4,069,450	37,426,507	6,802,801	4,069,450
influenza	154,808,555	15	1,957,370	4,259,746	3,304,035	4,477,322
einstein.en	467,626,544	139	212,903	643,338	427,142	601,755
kernel	257,961,616	162	1,374,650	3,769,839	2,870,350	3,795,801

#### Results: self-indexes



## Results: locate operation



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- What other types of queries can we support using local consistency?

# Questions?