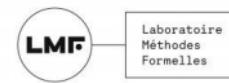


Tropical algebra for piecewise complexity

Isa Vialard

Joint work with Maëva Veron and Philippe Schnoebelen



MAX PLANCK INSTITUTE
FOR SOFTWARE SYSTEMS

♣ Subword relation

RADAR \preceq *ABRACADABRA*

◆ Simon's congruence

$u \equiv_m v$ iff $\forall w$ s.t. $|w| \leq m$, $w \preceq u \iff w \preceq v$

Ex. *NATIONALIST* \equiv_2 *ANTINATIONALIST*,

but *NATIONALIST* $\not\equiv_3$ *ANTINATIONALIST*: *INO* is a distinguisher.

Definition (Piecewise complexity)

$h(u)$ is the smallest k such that $u \equiv_k v$ implies $u = v$.

♣ Example: $h(abba) = 3$

$abba$ is the only word u such that $aba, bb \preceq u$ but $bab, aaa, bbb \not\preceq u$

¹Sakarovitch J., Simon I. Subwords. Combinatorics on Words.

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Definition (Distance¹)

$$\begin{aligned}\delta(u, v) &= \max\{k : u \equiv_k v\} \\ &= |s| - 1 \text{ for } s \text{ a shortest distinguisher.}\end{aligned}$$

♣ Example: $\delta(abba, baba) = 2$

baa is a shortest distinguisher so $abba \equiv_2 baba$

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Definition (Distance¹)

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♦ Not intuitive

$$\delta(aaaa, b) = 0 \text{ but } \delta(aaaa, aaaaa) = 6$$

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$h = \max$ of distance

Definition (Piecewise complexity)

$h(u)$ is the smallest k such that $u \equiv_k v$ implies $u = v$.

Definition (Distance²)

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◆ Theorem

$$h(u) = \max_{v \neq u} \delta(u, v) + 1$$

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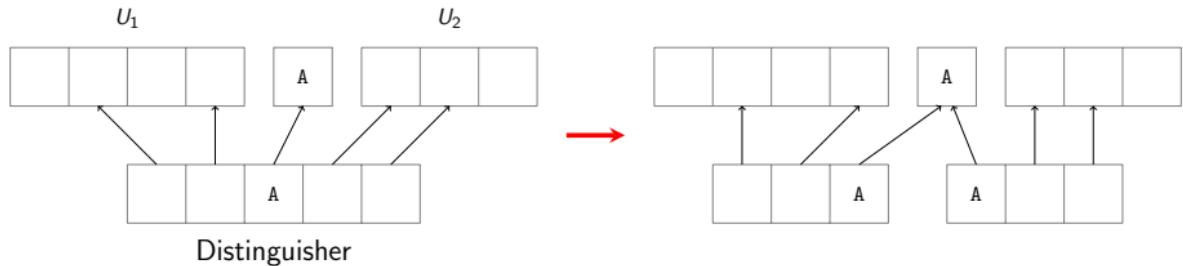
$$\begin{aligned}h(u) &= \max_{v \neq u} \delta(u, v) + 1 \\ &= \max_{\substack{u=u_1u_2 \\ a \in \Sigma}} \delta(u, u_1 \mathbf{a} u_2) + 1\end{aligned}$$

$h = \text{max of distance}$

$$h(u) = \max_{\substack{u=u_1u_2 \\ a \in \Sigma}} \delta(u, u_1 \textcolor{red}{a} u_2) + 1$$

and

$$\begin{aligned} \delta(u, u_1 \textcolor{red}{a} u_2) &= \delta(u_1, u_1 \textcolor{red}{a}) + \delta(u_2, \textcolor{red}{a} u_2) \\ &= r(u_1, \textcolor{red}{a}) + \ell(\textcolor{red}{a}, u_2)^3 \end{aligned}$$



³Simon, I. Piecewise testable events. In Automata Theory and Formal Languages (1975)

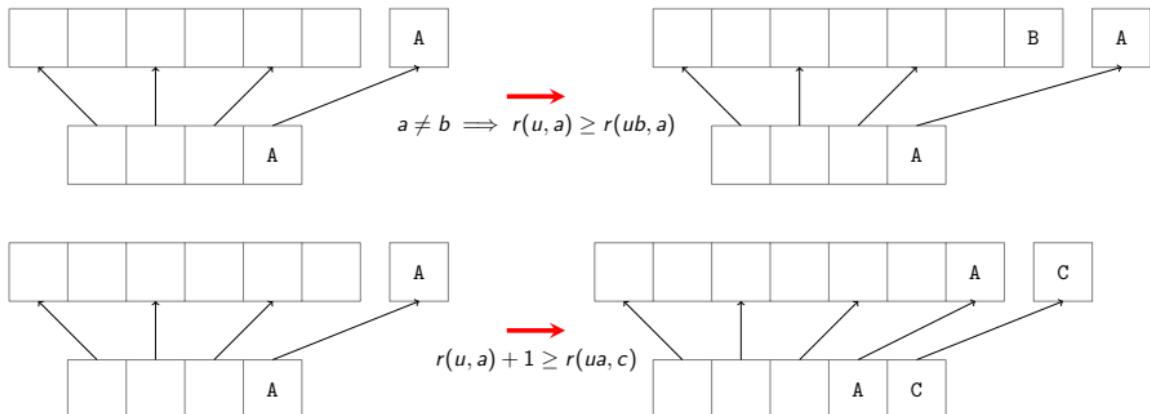
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Computing r

◆ Theorem ⁴

$$r(\epsilon, a) = 0$$

$$r(ub, a) = \begin{cases} r(u, a) + 1 & \text{if } a = b, \\ \min(r(u, b) + 1, r(u, a)) & \text{if } a \neq b. \end{cases}$$



Computing r

◆ Theorem 5

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♣ Example

Let $u = \text{ABBACCBCCABAABC}$ over $A = \{\text{A, B, C}\}$. The r -table of u is

w	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
$r(i, \text{A})$	0	1	1	1	2	1	1	1	1	1	2	2	3	4	4	3
$r(i, \text{B})$	0	0	1	2	2	1	1	2	2	2	2	3	3	3	4	3
$r(i, \text{C})$	0	0	0	0	0	1	2	2	3	4	2	2	2	2	2	3

⁵Schnoebelen, Vialard, On the piecewise complexity of words. Acta Informatica (2025) 7/22

Computing h from r and ℓ

Definition

The r -table R_u : $R_u(i, j) = r(u_j, a_i)$ where $u_j = u[1, j]$

The ℓ -table L_u : $L_u(i, j) = \ell(a_i, u'_j)$ where $u'_j = u[j + 1, |u|]$

Example

Here are R_u and L_u for $u = \text{CBBCAC}$, assuming $\Sigma = \{\text{A, B, C}\}$:

$$\begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & 2 & 1 & 1 \\ 0 & 1 & 1 & 1 & 2 & 1 & 2 \end{pmatrix} \quad \begin{matrix} r(u[4], \text{C}) = 2 \\ \ell(\text{C}, u[4,]) = 1 \end{matrix} \quad \begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 0 & 0 \\ 2 & 2 & 1 & 0 & 0 & 0 & 0 \\ 2 & 1 & 1 & 2 & 1 & 1 & 0 \end{pmatrix}$$

◆ Scanning $(R + L)_u$ for maximum: Complexity in $|u| \cdot |\Sigma|^6$

⁶Schnoebelen, Vialard, On the piecewise complexity of words. Acta Informatica (2025)

Our motivation

Definition

SLP = deterministic production rules $X \rightarrow YZ, X \rightarrow A$, generating 1 word

◆ Goal: compute $h(u)$ for u given as a SLP

... in size of SLP (instead of size of u)

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♣ Unfortunately $h(uv) \neq h(u) + h(v)$

Ex: $h(ab) = h(ba) = 2$ but $h(abba) = 3$

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Convexity: $h(u), h(v) \leq h(uv) \leq h(u) + h(v) - 1$

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Convexity: $h(u), h(v) \leq h(uv) \leq h(u) + h(v) - 1$

We need a *signature* $S(u)$

- with enough information to compute $h(u)$
- such that $S(uv)$ can be computed from $S(u)$ and $S(v)$

Now tropical algebra appears!

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Let $u = \text{ABBACCBCCABAABC}$ over $A = \{\text{A, B, C}\}$. The r -table of u is

w	i	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
		A	B	B	A	C	C	B	C	C	A	B	A	A	B	C	
$r(i, \text{A})$		0	1	1	1	2	1	1	1	1	1	2	2	3	4	4	3
$r(i, \text{B})$		0	0	1	2	2	1	1	2	2	2	2	3	3	3	4	3
$r(i, \text{C})$		0	0	0	0	0	1	2	2	2	3	4	2	2	2	2	3

$$\begin{pmatrix} 2 & 2 & 2 \end{pmatrix} = \begin{pmatrix} 1+1 \\ \min 2, 1+1 \\ \min 4, 1+1 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 4 \end{pmatrix} \cdot \begin{pmatrix} 1 & 1 & 1 \\ * & 0 & * \\ * & * & 0 \end{pmatrix}_{M_A}$$

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◆ Min-plus algebra

- Just like Max-plus
- $*$ denotes $+\infty$

$$\begin{pmatrix} 2 & 2 & 2 \end{pmatrix} = \begin{pmatrix} 1+1 \\ \min 2, 1+1 \\ \min 4, 1+1 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 4 \end{pmatrix} \cdot \begin{pmatrix} 1 & 1 & 1 \\ * & 0 & * \\ * & * & 0 \end{pmatrix}_{M_A}$$

♣ Introduced by Simon

- $\exists m?$ s.t. $A^m = A^*$
- Finding minimal paths in graph

Now tropical algebra appears!

◆ Min-plus algebra

- Just like Max-plus
- $*$ denotes $+\infty$

Definition

For $u = u_1 \dots u_m$, $M_u = M_{u_1} \cdot \dots \cdot M_{u_n}$.

$$\begin{pmatrix} 2 & 2 & 2 \end{pmatrix} = \begin{pmatrix} 1+1 \\ \min 2, 1+1 \\ \min 4, 1+1 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 4 \end{pmatrix} \cdot \begin{pmatrix} 1 & 1 & 1 \\ * & 0 & * \\ * & * & 0 \end{pmatrix}_{M_A}$$
$$= \begin{pmatrix} 0 & 0 & 0 \end{pmatrix} \cdot M_u$$

w	i	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
$r(i, A)$	0	1	1	1	2	1	1	1	1	1	2	2	3	4	4	3	
$r(i, B)$	0	0	1	2	2	1	1	2	2	2	2	3	3	3	4	3	
$r(i, C)$	0	0	0	0	0	1	2	2	3	4	2	2	2	2	2	3	

Expanding the r -table

Definition

The matrix r -table F_u :

$$r(i, a_j) = (0 \dots 0) \cdot M_{u_i}(\cdot, j) = (0 \dots 0) \cdot F_u(i, j)$$

♣ Example ($u = \text{CBBCAC}$)

$$\begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & 2 & 1 & 1 \\ 0 & 1 & 1 & 1 & 2 & 1 & 2 \end{pmatrix} \rightarrow \begin{vmatrix} \begin{pmatrix} 0 \\ * \\ * \end{pmatrix} & \begin{pmatrix} 0 \\ * \\ 1 \end{pmatrix} & \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} & \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} & \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} & \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix} & \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix} \\ \begin{pmatrix} * \\ 0 \\ * \end{pmatrix} & \begin{pmatrix} * \\ 0 \\ 1 \end{pmatrix} & \begin{pmatrix} * \\ 1 \\ 2 \end{pmatrix} & \begin{pmatrix} * \\ 2 \\ 3 \end{pmatrix} & \begin{pmatrix} * \\ 2 \\ 2 \end{pmatrix} & \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix} & \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix} \\ \begin{pmatrix} * \\ * \\ 0 \end{pmatrix} & \begin{pmatrix} * \\ * \\ 1 \end{pmatrix} & \begin{pmatrix} * \\ 1 \\ 1 \end{pmatrix} & \begin{pmatrix} * \\ 1 \\ 1 \end{pmatrix} & \begin{pmatrix} * \\ 1 \\ 2 \end{pmatrix} & \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix} & \begin{pmatrix} 2 \\ 3 \\ 3 \end{pmatrix} \end{vmatrix}$$

Now combining the new r and ℓ -tables

♣ Example ($u = \text{CBBCAC}$)

$$\begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & 2 & 1 & 1 \\ 0 & 1 & 1 & 1 & 2 & 1 & 2 \end{pmatrix}$$

$$r(u[4], C) = 2$$

$$\ell(C, u[4,]) = 1$$

$$\begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 0 & 0 \\ 2 & 2 & 1 & 0 & 0 & 0 & 0 \\ 2 & 1 & 1 & 2 & 1 & 1 & 0 \end{pmatrix}$$

$$\left| \begin{pmatrix} 0 \\ * \end{pmatrix} \begin{pmatrix} 0 \\ * \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right|$$

$$\left| \begin{pmatrix} 1 \\ 2 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 \\ * \end{pmatrix} \right|$$

$$(*\ 2\ 2) \otimes (1\ * \ 1) = \begin{pmatrix} * & * & * \\ 3 & * & 3 \\ 3 & * & 3 \end{pmatrix}$$

Our signature

◆ Signature $S(u)$

- M_u, M_{u^R}
- H_u the maximum elements of $F_u \otimes G_u$

♣ Compute $h(u)$ from $S(u)$

- $r(u_j, a_i) = (0 \dots 0) \cdot F_u(i, j) = \min F_u(i, j)$
- $r(u_j, a_i) + \ell(a_i, u') = \min(F_u \otimes G_u)(i, j)$ for $u = u_j u'$

Hence

$$\begin{aligned} \max_{i,j} r(u_j, a_i) + \ell(a_i, u') + 1 &= \max_{i,j} \min(F_u \otimes G_u)(i, j) + 1 \\ &= \max_{M \in H_u} \min(F_u \otimes G_u)(i, j) + 1 \end{aligned}$$

Concatenating tables

♣ Example: F_{uv} for $u = \text{CBB}$, $v = \text{CAC}$

$$\begin{array}{c} \begin{array}{ccccccccc} & \text{C} & \text{B} & \text{B} & & \text{C} & \text{A} & \text{C} \\ \left| \begin{pmatrix} 0 \\ * \\ * \end{pmatrix} \begin{pmatrix} 0 \\ * \\ 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right| & :: & \left| \begin{pmatrix} 0 \\ * \\ * \end{pmatrix} \begin{pmatrix} 0 \\ * \\ 1 \end{pmatrix} \begin{pmatrix} 1 \\ * \\ 2 \end{pmatrix} \begin{pmatrix} 1 \\ * \\ 2 \end{pmatrix} \right| \\ \left| \begin{pmatrix} 0 \\ * \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} \begin{pmatrix} 2 \\ 3 \end{pmatrix} \right| & :: & \left| \begin{pmatrix} 0 \\ * \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \right| \\ \left| \begin{pmatrix} * \\ 0 \end{pmatrix} \begin{pmatrix} * \\ 1 \end{pmatrix} \begin{pmatrix} * \\ 1 \end{pmatrix} \begin{pmatrix} * \\ 1 \end{pmatrix} \right| & :: & \left| \begin{pmatrix} * \\ 0 \end{pmatrix} \begin{pmatrix} * \\ 1 \end{pmatrix} \begin{pmatrix} 1 \\ * \\ 1 \end{pmatrix} \begin{pmatrix} 2 \\ * \\ 2 \end{pmatrix} \right| \end{array} \\ \\ \begin{array}{c} \begin{array}{ccccccccc} \left(\begin{pmatrix} 0 \\ * \\ * \end{pmatrix} \begin{pmatrix} 0 \\ * \\ 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} M_u \begin{pmatrix} 0 \\ * \\ 1 \end{pmatrix} M_u \begin{pmatrix} 1 \\ * \\ 2 \end{pmatrix} M_u \begin{pmatrix} 1 \\ * \\ 2 \end{pmatrix} \right) \\ \left(\begin{pmatrix} * \\ 0 \\ * \end{pmatrix} \begin{pmatrix} * \\ 0 \\ 1 \end{pmatrix} \begin{pmatrix} * \\ 1 \\ 2 \end{pmatrix} \begin{pmatrix} * \\ 2 \\ 3 \end{pmatrix} M_u \begin{pmatrix} * \\ 0 \\ 1 \end{pmatrix} M_u \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} M_u \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \right) \\ \left(\begin{pmatrix} * \\ 0 \\ 0 \end{pmatrix} \begin{pmatrix} * \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} * \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} * \\ 1 \\ 1 \end{pmatrix} M_u \begin{pmatrix} * \\ 1 \\ 1 \end{pmatrix} M_u \begin{pmatrix} 1 \\ * \\ 1 \end{pmatrix} M_u \begin{pmatrix} 2 \\ * \\ 2 \end{pmatrix} \right) \end{array} \\ \begin{array}{ccccccccc} \text{C} & \text{B} & \text{B} & \text{C} & \text{A} & \text{C} \\ \hline \end{array} \end{array} \end{array}$$

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$$\left| \begin{array}{cccc} \text{C} & \text{B} & \text{B} & \\ \left(\begin{matrix} 0 \\ * \\ * \end{matrix} \right) & \left(\begin{matrix} 0 \\ * \\ 1 \end{matrix} \right) & \left(\begin{matrix} 0 \\ 1 \\ 1 \end{matrix} \right) & \left(\begin{matrix} 0 \\ 1 \\ 1 \end{matrix} \right) \\ \left(\begin{matrix} * \\ 0 \\ * \end{matrix} \right) & \left(\begin{matrix} * \\ 0 \\ 1 \end{matrix} \right) & \left(\begin{matrix} * \\ 1 \\ 2 \end{matrix} \right) & \left(\begin{matrix} * \\ 2 \\ 3 \end{matrix} \right) \\ \left(\begin{matrix} * \\ * \\ 0 \end{matrix} \right) & \left(\begin{matrix} * \\ * \\ 1 \end{matrix} \right) & \left(\begin{matrix} * \\ 1 \\ 1 \end{matrix} \right) & \end{array} \right| :: \left| \begin{array}{ccc} \text{C} & \text{A} & \text{C} \\ \left(\begin{matrix} 0 \\ * \\ * \end{matrix} \right) & \left(\begin{matrix} 0 \\ * \\ 1 \end{matrix} \right) & \left(\begin{matrix} 1 \\ * \\ 2 \end{matrix} \right) \\ \left(\begin{matrix} * \\ 0 \\ * \end{matrix} \right) & \left(\begin{matrix} * \\ 0 \\ 1 \end{matrix} \right) & \left(\begin{matrix} 1 \\ 0 \\ 1 \end{matrix} \right) \\ \left(\begin{matrix} * \\ * \\ 0 \end{matrix} \right) & \left(\begin{matrix} * \\ * \\ 1 \end{matrix} \right) & \left(\begin{matrix} 1 \\ * \\ 2 \end{matrix} \right) \end{array} \right|$$

$$= \left| \begin{array}{ccccccccc} \text{C} & \text{B} & \text{B} & & \text{C} & & \text{A} & & \text{C} \\ \left(\begin{matrix} 0 \\ * \\ * \end{matrix} \right) & \left(\begin{matrix} 0 \\ * \\ 1 \end{matrix} \right) & \left(\begin{matrix} 0 \\ 1 \\ 1 \end{matrix} \right) & \left(\begin{matrix} 0 \\ 1 \\ 1 \end{matrix} \right) & M_u \left(\begin{matrix} 0 \\ * \\ 1 \end{matrix} \right) & M_u \left(\begin{matrix} 1 \\ * \\ 2 \end{matrix} \right) & M_u \left(\begin{matrix} 1 \\ * \\ 2 \end{matrix} \right) \\ \left(\begin{matrix} * \\ 0 \\ * \end{matrix} \right) & \left(\begin{matrix} * \\ 0 \\ 1 \end{matrix} \right) & \left(\begin{matrix} * \\ 1 \\ 2 \end{matrix} \right) & \left(\begin{matrix} * \\ 2 \\ 3 \end{matrix} \right) & M_u \left(\begin{matrix} * \\ 0 \\ 1 \end{matrix} \right) & M_u \left(\begin{matrix} 1 \\ 0 \\ 1 \end{matrix} \right) & M_u \left(\begin{matrix} 1 \\ 0 \\ 1 \end{matrix} \right) \\ \left(\begin{matrix} * \\ * \\ 0 \end{matrix} \right) & \left(\begin{matrix} * \\ * \\ 1 \end{matrix} \right) & \left(\begin{matrix} * \\ 1 \\ 1 \end{matrix} \right) & \left(\begin{matrix} * \\ 1 \\ 1 \end{matrix} \right) & M_u \left(\begin{matrix} * \\ * \\ 1 \end{matrix} \right) & M_u \left(\begin{matrix} 1 \\ * \\ 1 \end{matrix} \right) & M_u \left(\begin{matrix} 2 \\ * \\ 2 \end{matrix} \right) \end{array} \right|$$

Concatenating tables

♦ Signature $S(u)$

- M_u, M_{u^R}
- H_u the maximum elements of $F_u \otimes G_u$

♣ Compute $S(uv)$ from $S(u)$ and $S(v)$

- $F_{uv}(i, j) = F_u(i, j) \text{ or } M_u \cdot F_v(i, j)$
- $G_{uv}(i, j) = G_u(i, j) \cdot M_{v^R} \text{ or } F_v(i, j)$

Hence $H_{uv} \subseteq M_u \cdot H_v \cup H_u \cdot M_{v^R}$

Size of our signature

♦ Signature $S(u)$

- M_u, M_{u^R}
- H_u the set of maximum elements of $F_u \otimes G_u$

♣ Does $|H_u|$ explodes when concatenating?

Size of our signature

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NO: $H_u \leq |\Sigma| \cdot (2|\Sigma| + 1)$

- Bound on the largest antichain if $F_u \otimes G_u$

Size of our signature

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♣ Does $|H_u|$ explodes when concatenating?

NO: $H_u \leq |\Sigma| \cdot (2|\Sigma| + 1)$

- Bound on the largest antichain if $F_u \otimes G_u$
- Almost tight: $|H_{u_k}| \geq k \cdot (k + 1)$ for

$$u_k = a_1 \dots a_k a_k a_{k-1} a_k \dots a_2 \dots a_k a_1 a_k \dots a_2 \dots a_k a_{k-1} a_k a_k \dots a_1$$

Other use of tropical algebra

Definition

Minimality index $\rho(u)$ is the smallest k such that $u \equiv_k v$ implies $u = v \dots$

... for all $v \preceq u$

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◆ Theorem

- $\rho(u) = \max_{u=u_1 \textcolor{red}{\bullet} u_2} \delta(u, u_1 u_2) + 1$
- $\rho(u) \leq h(u) - 1$ ($=$ if $|\Sigma| = 2$)

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- $\rho(u) \leq h(u) - 1$ ($=$ if $|\Sigma| = 2$)

♣ Tweak our algorithm for ρ :

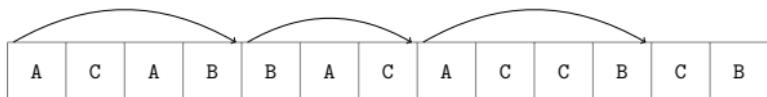
Bounds on $|H_u|$?

- $|H_u| \leq |\Sigma| \cdot (|\Sigma| + 2)$.
- Might not be tight

Other use of tropical algebra

Definition

Universality index $\iota(u) = \max\{k: |v| \leq k \implies v \preceq u\}$



Ex: $\iota(u) = 3$

♣ Example ($u = \text{ACABBACACCBBCB}$)

Theorem

$$\iota(u) = \min M_u$$

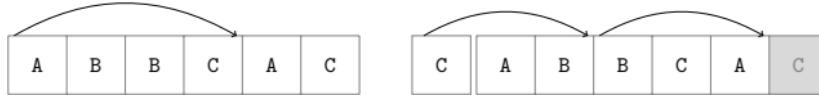
$$M_u = \begin{pmatrix} 4 & 5 & 5 \\ 3 & 4 & 4 \\ 3 & 4 & 4 \end{pmatrix}$$

Other use of tropical algebra

Definition

Circular universality index

$$\zeta(u) = \max_{u=u_1u_2} \iota(u_2u_1)$$



Ex: $\iota(u) = 1$ but $\zeta(u) = 2$

Other use of tropical algebra

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Circular universality index

$$\zeta(u) = \max_{u=u_1u_2} \iota(u_2u_1)$$



Ex: $\iota(u) = 1$ but $\zeta(u) = 2$

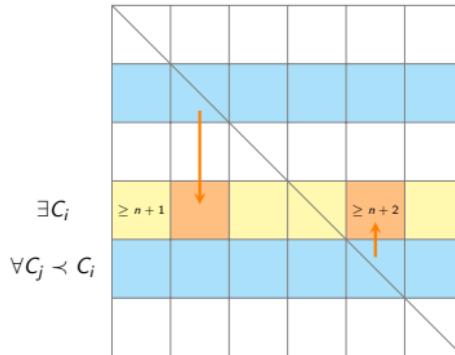
♣ Can you see the pattern ?

$$M_{\text{ABBCAC}} = \begin{pmatrix} 2 & 2 & 3 \\ 2 & 2 & 3 \\ 2 & 1 & 2 \end{pmatrix}$$

◆ Theorem

If $\iota(u) = n$, then $\zeta(u) = n + 1$ iff:

Otherwise $\zeta(u) = n$



♣ Can't guess this pattern ⁷

$$M_{\text{ABBCAC}} = \begin{pmatrix} 2 & 2 & 3 \\ 2 & 2 & 3 \\ 2 & 1 & 2 \end{pmatrix}$$

⁷Schnoebelen, Veron,(2023). On Arch Factorization and Subword Universality for Words and Compressed Words, WORDS 2023.

Conclusion

♣ Tropical algebra is a natural tool to study these kind of measures

- Easy to implement data structures
- M_u is a gold mine: we can extract $\iota(u), \zeta(u), \dots$
- ... and how u “behaves” when concatenated to other words

♦ Some open questions

- Tight bounds on H_u for $\rho(u)$
- Bounds on the relation between $|u|$ and $h(u)$
- Same question for $\rho(u)$