

2.1.7 Lex' theorem

2.1 Lex and shortlex orders, Hankel matrix

We define a linear order \leq over the alphabet Σ ,
 typical examples: $\Sigma = \{a, b\}$, $a < b$

$$\Sigma = \{a_1, \dots, a_n\}, \quad a_1 < a_2 < \dots < a_n$$

We define the lexical order over Σ^* in the following way:
 for $u, v \in \Sigma^*$, $u \leq_{\text{lex}} v$ if \bullet u prefix of v ($u \cdot u' = v$ for some u')

$$\bullet \text{ for some } u_0, u_1, v_1, a < b, \\ u = u_0 \cdot a \cdot u_1 \\ v = u_0 \cdot b \cdot v_1$$

\leq_{lex} is a linear order

We define the shortlex order in the following way:

$$u \leq_{\text{sllex}} v \text{ if } \bullet |u| < |v|$$

$$\bullet |u| = |v| \text{ and } u \leq_{\text{lex}} v$$

$$\epsilon <_{\text{sllex}} a <_{\text{sllex}} b <_{\text{sllex}} aa <_{\text{sllex}} ab <_{\text{sllex}} ba <_{\text{sllex}} bb <_{\text{sllex}} aaa <_{\text{sllex}} \dots$$

Let $s \in R \langle \langle \Sigma \rangle \rangle$. The Hankel matrix of s is defined

as follow: $H_s(u, v) = s(u \cdot v)$ (infinite)

$$\text{Fact: } H_s(u \cdot w, v) = H_s(u, v \cdot w)$$

ex: $\Sigma = \{a, b\}$

$H_{\text{length}} :=$

	ϵ	a	b	aa	ab	ba	bb	aaa	\dots
ϵ	0	1	1	2	2	2	2	3	\dots
a	1	2	2	3	3	3	3	4	\dots
b	1	2	2	3	3	3	3	4	\dots
aa	2	3	3	4	4	4	4	5	\dots
ab	2	3	3	4	4	4	4	5	\dots
ba	2	3	3	4	4	4	4	5	\dots
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots

Statement and proof

Now, we suppose that R is actually a field \mathbb{F} (x is commutative and $+$, \times have inverses)

We have a notion of linearly independence (of columns, lines...)
 of a notion of ranks (rows)

Theorem [Fliess, 1974]

$s \in \mathbb{F}\langle\langle \Sigma \rangle\rangle$ is rational iff H_s has finite rank.
 Moreover, in that case, the finite rank r gives the minimal size of a weighted automaton recognising s .

proof

• Suppose s recognised by $A = \langle N, \lambda, (\delta a)_a, \mu \rangle$
 for each $i \in N$, and we define, for $w \in \Sigma^*$ and p
 a run $q_0 \dots q_n$ ending on i (meaning $q_{n+1} = i$), we define
 the left weight of p as $L_i(p) = \lambda(q_0) \times \text{transweight}(p) \times \mu(q_n)$ (meaning w/out $p(q_n)$)
 $\text{Lweight}(p)$

and $L_i(u)$ is the sum $\sum_{\substack{p \text{ run via } u \\ \text{ending on } i}} \text{Lweight}(p)$.

L_i is the infinite ^{column} line $[L_i(\epsilon) \ L_i(a) \ \dots]$

Similarly, we have R_i as the infinite ^{column} line
 where $R_i(v)$ is defined symmetrically (right weight)

Then, for every all $u, v \in \Sigma^*$, we have

$$H_s(u, v) = \sum_{i \in N} L_i(u) \times R_i(v)$$

$H_s = \sum_{i \in N} L_i \times R_i$, of rank $\leq N$ (every $L_i \times R_i$ has rank at most 1)

• suppose now that H_s has finite rank N
 $(L_{u_i})_{i < N}$ base of the matrix.

i) ~~for every~~ exist $\lambda(0), \lambda(1), \dots, \lambda(N-1)$ such that

$$L_\varepsilon = \sum_{i < N} \lambda(i) \cdot L_{u_i}$$

ii) for every $i < N$, $a \in \Sigma$, exists

$$\delta_a(i, 0), \delta_a(i, 1), \dots, \delta_a(i, N-1)$$

such that

$$L_{u_i \cdot a} = \sum_{j < N} \delta_a(i, j) L_{u_j}$$

iii) finally, define $\mu(i) = L_{u_i}(\varepsilon)$

$\langle N, \lambda, (\delta_a)_a, \mu \rangle$ ~~defines~~ recognises s !

proof by example:

$$s(a \cdot b) = H_s(\varepsilon, a \cdot b)$$

$$= L_\varepsilon(ab)$$

$$= \sum_{i < N} \lambda(i) L_{u_i}(ab)$$

$$\xrightarrow{H_s(u_i, ab)}$$

$$= \sum_{i < N} \lambda(i) L_{u_i \cdot a}(b)$$

$$\xrightarrow{H_s(u_i \cdot a, b)}$$

$$= \sum_{i < N} \lambda(i) \sum_{j < N} \delta_a(i, j) L_{u_j \cdot b}(\varepsilon)$$

$$= \sum_{i < N} \lambda(i) \sum_{j < N} \delta_a(i, j) \sum_{k < N} \delta_b(j, k) L_{u_k}(\varepsilon)$$

$$= \sum_{i < N} \lambda(i) \sum_{j < N} \delta_a(i, j) \sum_{k < N} \delta_b(j, k) \mu(k)$$

$$= \text{weight}_A(a \cdot b) \quad // \text{the value of } A \text{ in terms of matrices}$$

□