Weighted Automata

Weighted Automata and Monadic Second Order Logic

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Revised and expanded slides originally presented at

GandALF 2013

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Probabilistic automata (Rabin 1961)

A vector $\alpha = (\alpha_1, \ldots, \alpha_r) \in \mathbb{R}^r$ is stochastic if each $\alpha_i \ge 0$ and $\sum_i \alpha_i = 1$.

A matrix $\mu \in \mathbb{R}^{r \times r}$ is row-stochastic (column-stochasttic) if each row-vector (column-vector) is stochastic. μ is doubly stochastic if it is both row- and column-stochastic.

A Probabilistic Automaton (PA) A of size r is given by:

- A set $\{\mu_{\sigma} : \sigma \in \Sigma\}$ of $r \times r$ doubly stochastic matrices;
- Two stochastic vectors $\lambda, \gamma \in \mathcal{F}^r$.
- A defines a function $f_A : \Sigma^* \to \mathbb{R}$

$$f_A(w) = f_A(\sigma_1 \circ \sigma_2 \circ \ldots \circ \sigma_n) = \lambda \mu_{\sigma_1} \mu_{\sigma_2} \cdot \ldots \cdot \mu_{\sigma_n} \gamma^t$$

• A function $f: \Sigma^* \to \mathbb{R}$ is PA-recognizable if $f = f_A$ for some PA A.

Intuition behind probabilistic automata

- The automaton has *r* states.
- λ gives the probability λ_i that the automaton is in state i when reading the empty word.
- μ_{σ} is the transition matrix for the transition when reading $\sigma_{...}$
- γ gives the probability γ_i that state *i* is an accepting state.

Computability and Definability (236331), December 2015 Weighted Automata

Multiplicity automata (Schutzenberger, 1961)

A Multiplicity Automaton (MA) A of size r over a field \mathcal{F} is given by:

- A set $\{\mu_{\sigma} : \sigma \in \Sigma\}$ of $r \times r$ matrices over \mathcal{F} ;
- Two vectors $\lambda, \gamma \in \mathcal{F}^r$.

• A defines a function
$$f_A : \Sigma^* \to \mathcal{F}$$

$$f_A(w) = f_A(\sigma_1 \circ \sigma_2 \circ \ldots \circ \sigma_n) = \lambda \mu_{\sigma_1} \mu_{\sigma_2} \cdot \ldots \cdot \mu_{\sigma_n} \gamma^t$$

• A function $f: \Sigma^* \to \mathcal{F}$ is MA-recognizable if $f = f_A$ for some MA A.

Probabilistic automata (PA) and Multiplicity automata (MA) where introduced independently, generalizing the developments described in the famous paper by M. Rabin and D. Scott (1959).

Word functions and power series

Let \mathcal{F} be a field (or semi-ring) and Σ an alphabet.

We can view Σ as a set of non-commutative indeterminates and Σ^* is its set of monomials.

A function $f: \Sigma^* \to \mathcal{F}$ the defines a power series

$$S_f(w) = \sum_{w \in \Sigma^*} f(w)w$$

A power series is rational if it can be obtained from polynomials by addition, multiplication, external products and the star-operation.

Regular languages and power series

We define a language $L(f) = \{w \in \Sigma^* : f(w) \neq 0\}.$

L(f) is FA-recognizable if there is a deterministic finite automaton A which accepts L(f).

Theorem: (Kleene-Schützenberger)

In the case of $\mathcal{F} = \mathbb{Z}_2$ the following are equivalent:

- (i) L(f) is FA-recognizable;
- (ii) L(f) is regular;

(iii) $S_f(w)$ is rational.

MA-Recognizable word functions

A function $f: \Sigma^* \to \mathcal{F}$ is MA-recognizable if there exists an MA A such that $f_A = f$.

Theorem: (Schützenberger 1961)

For arbitrary semi-rings \mathcal{F} the following are equivalent:

- (i) f MA-recognizable
- (ii) $S_f(w)$ is rational

Is there an analogue for regular expressions for MA over \mathcal{F} ?

Hankel matrices in classical mathematics

(over a field \mathcal{F})

Let $f : \mathcal{F} \to \mathcal{F}$ be a function.

A finite or infinite matrix $H(f) = h_{i,j}$ over a field \mathcal{F} is a Hankel matrix for f if $h_{i,j} = f(i+j)$.

Hankel matrices have many applications in: numeric analysis, probability theory and combinatorics.

- Padé approximations
- Orthogonal polynomials
- Probability theory (theory of moments)
- Coding theory (BCH codes, Berlekamp-Massey algorithm)
- Combinatorial enumerations (Lattice paths, Young tableaux, matching theory)

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Hankel matrices over words

Let Σ be a finite alphabet and \mathcal{F} be a field and let $f: \Sigma^* \to \mathcal{F}$ be a function on words.

A finite or infinite matrix $H(f) = h_{u,v}$ indexed over the words $u, v \in \Sigma^*$ is a Hankel matrix for f if $h_{u,v} = f(u \circ v)$. Here \circ denotes concatenation.

Hankel matrices over words have applications in

- Formal language theory and stochastic automata, J. Carlyle and A. Paz 1971
- Learning theory (exact learning of queries).
 A.Beimel, F. Bergadano, N. Bshouty, E. Kushilevitz, S. Varricchio 1998
 J. Oncina 2008
- Definability of picture languages.
 O. Matz 1998, and D. Giammarresi and A. Restivo 2008

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Hankel matrices for graphs

If we want to define Hankel matrices for (labeled) graphs,

what plays the role of concatenation?

- Disjoint union Used by Freedman, Lovász and Schrijver, 2007, for characterizing multiplicative graph parameters over the real numbers
- k-unions (connections, connection matrices)
 Used by Freedman, Lovász, Schrijver and Szegedy, 2007ff, for characterizing various forms and partition functions.
- Joins, cartesian products, generalized sum-like operations used by Godlin, Kotek and JAM to prove non-definability.

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Multiplicity Automata and Hankel matrices (over a field)

THEOREM: (J. Carlyle and A. Paz 1971)

For a function $f: \Sigma^* \to \mathcal{F}$ the following are equivalent:

- (i) f is MA-recognizable;
- (ii) S_f is rational
- (iii) the Hankel matrix H(f) has finite rank over \mathcal{F} .

This is an **ALGEBRAIC characterization of MA-recognizability**.

Weighted Automata

Proof of the Carlyle-Paz Theorem, I

The following proof is taken from

 Beimel, A., Bergadano, F., Bshouty, N. H., Kushilevitz, E., and Varricchio, S. Learning functions represented as multiplicity automata. Journal of the ACM, 47(3), (2000), pp. 506-530.

There are some changes in notation:

- We have matrices ν_{σ} for each $\sigma \in \Sigma$.
- λ is not used. Instead, $\nu(\epsilon) = \mathbb{W}$, and $f_A(w) = [\nu(w)]_1 \cdot \gamma$ with $w = \sigma_1 \sigma_2 \dots \sigma_n$, and $\nu(w) = \nu_{\sigma_1} \cdot \nu_{\sigma_2} \dots \cdot \nu_{\sigma_n}$, where $[\nu]_1(w)$ is the first row of $\nu(w)$.
- In our notation this would be

$$f_A(w) = \lambda \cdot \mu(w) \cdot \gamma = [\nu(w)]_1 \cdot \gamma$$

hence

$$\lambda \cdot \mu(w) = [\nu(w)]_1$$

Proof of the Carlyle-Paz Theorem, II

Let $f: \Sigma^* \to \mathcal{F}$ be recognized by a multiplicity automaton A with r states.

Let H(f) be the Hankel matrix of f.

Claim: The infinite matrix r(H(f)) has finite rank $\leq r$.

- Let R, C be matrices where rows of R (columns of C) are indexed by Σ^* and columns of R (rows of C) are indexed by [r].
- The (v, i)-entry of R is $[\nu(v)]_{1,i} = \lambda \cdot \nu(v)_i$.
- The (i, w)-entry of R is $[\nu(w)]_i \cdot \gamma$.

Proof of the Carlyle-Paz Theorem, III

Now we compute the rank of H(f).

$$[H(f)]_{v,w}f(v \circ w) = f_A(v \circ w) = [\nu(v \circ w)]_1 \cdot \gamma$$
$$= [\nu(v) \cdot \nu(w)]_1 \cdot \gamma = \sum_{i=1}^r [\nu(v)]_{1,i} \cdot [\nu(w)]_i \cdot \gamma = R_v \cdot C^w$$

Here R_v is the *v*-row of *R* and C^w is *w*-column of *C*.

From Linear Algebra we (should) know:

- The ranks of R and C are bounded by r.
- The rank of $R \cdot C \leq \min\{rk(R), rk(C)\}$.

Hence, the rank of $H(f) \leq \min\{rk(R), rk(C)\}$.

Proof of the Carlyle-Paz Theorem, IV

Now let f be given with H(f) = H and rk(H) = r.

We want to construct an automaton $A = A_f$ recognizing f.

- Let $B = \{H_{v_1}, \ldots, H_{v_r}\}$ be a basis for the rows of H.
- $\gamma = (f(v_1), \ldots, f(v_r)).$
- For $\sigma \in \Sigma$, let $H_{v_i \circ \sigma}$ be the row of $v_i \circ \sigma$ in H.
- We write $H_{v_i \circ \sigma}$ as a linear combination of vectors of B:

$$H_{v_i \circ \sigma} = \sum_{j=1}^r b_{i,j} \cdot H_{v_j}$$

• Now we put $[\nu_{\sigma}]_{i,j} = b_{i,j}$.

Proof of the Carlyle-Paz Theorem, V

Now we prove by induction on $\ell(w)$ that for all $i \leq r$ we have

 $[\nu(w)]_i \cdot \gamma = f(v_i \circ \sigma)$ hence, for $v_1 = \epsilon$, $f_A(w) = [\nu_1(w) \cdot \gamma = f(v_1 \circ w) = f(w).$

- Induction base: $w = \epsilon$. Here $\nu(\epsilon) = \mathbb{H}$ and $[\nu(\epsilon)]_i \cdot \gamma = \gamma_i = f(v_1) = f(v_1 \circ \epsilon)$.
- By definition, we have $[\nu_{\sigma}]_{i,j} = b_{i,j}$. Hence

$$f(v_1 \circ \sigma \circ w) = H_{v_1 \circ \sigma}(w) = \sum_{j=1}^r b_{i,j} \cdot H_{v_j}(w)$$

• Since $H_{v_j}(w) = f(v_j \circ w)$, we get by induction hypothesis:

$$H_{v_j}(w) = \sum_{j=1}^r [\nu_\sigma]_{i,j} \cdot [\nu(w)]_j \cdot \gamma = [\nu(\sigma) \cdot \nu(w)]_i \cdot \gamma = [\nu(\sigma \circ w)]_i \cdot \gamma$$

The Büchi-Elgot-Trakhtenbrot Theorem (around 1960)

A word w of size n over an alphabet Σ can be considered as a structure

$$\mathfrak{A}_w = \langle [n], \langle_{nat}, P_{\sigma}, (\sigma \in \Sigma) \rangle$$

where $P_{\sigma} : \sigma \in \Sigma$ is a partition of [n] into possibly empty sets.

THEOREM: (R. Büchi, C. Elgot and B. Trakhtenbrot)

The following are equivalent:

- (i) L is FA-recognizable;
- (ii) *L* is regular;
- (iii) The class $\{\mathfrak{A}_w : w \in L\}$ of structures is **definable in Monadic Second Order Logic**.

Is there an analogue for MA-recognizability ?

$\mathsf{MSOLEVAL}_{\mathcal{F}}$

 $MSOLEVAL_{\mathcal{F}}$ consists of those functions mapping relational structures into \mathcal{F} which are definable in Monadic Second Order Logic MSOL.

The functions in MSOLEVAL_{\mathcal{F}} are represented as terms associating with each τ -structure \mathcal{A} a polynomial $p(\mathcal{A}, \overline{X}) \in \mathcal{F}[\overline{X}]$.

Similarly, CMSOLEVAL $_{\mathcal{F}}$ is obtained by replacing MSOL by Monadic Second Order Logic with modular counting CMSOL.

MSOLEVAL $_{\mathcal{F}}$ is defined inductively:

- (i) monomials are products of constants in \mathcal{F} and indeterminates in \overline{X} and the product ranges over elements a of \mathcal{A} which satisfy an MSOL-formula $\phi(a)$.
- (ii) polynomials are then defined as sums of monomials where the sum ranges over unary relations $U \subset A$ satisfying an MSOL-formula $\psi(U)$.

 $MSOLEVAL_{\mathcal{F}}$ was first studied in a sequence of papers on graph polynomials by J.A.M. variably co-authored with B. Courcelle, B. Godlin, T. Kotek, U. Rotics, B. Zilber.

We proceed now by examples of word functions in MSOLEVAL.

Examples of word functions in MSOLEVAL, I

Let $\Sigma = \{0,1\}$ and $w \in \Sigma^*$ be represented by the structure

$$\mathcal{A}_w = \langle [\ell(w)], <, P_0, P_1 \rangle.$$

Counting occurrences:

(i) The function $\sharp_1(w)$ counts the number of occurences of 1 in a word w can be written as

$$\sharp_1(w) = \sum_{i \in [n]: P_1(i)} 1.$$

(ii) The polynomial $X^{\sharp_1(w)}$ can be written as

$$X^{\sharp_1(w)} = \prod_{i \in [n]: P_1(i)} X.$$

Examples of word functions in MSOLEVAL, II

Let L be a regular language defined by the MSOL-formula ϕ_L .

The polynomial

$$\sharp_L(w) = \sum_{u \in L: \exists v_1, v_2(w = v_1 \circ u \circ v_2)} X^{\ell(u)}$$

is the generating function of the number of (contiguous) occurrences of words $u \in L$ in a word w of size i.

It can also be written as

$$\sharp_L(w) = \sum_{U \subseteq [n]: \psi_L(U)} \prod_{i \in U} X,$$

where $\psi_L(U)$ says that U is an interval and ϕ_L^U , the relativization of ϕ_L to U holds.

Examples of word functions in MSOLEVAL, III

Let $int(w) = \sum_{i=0}^{\ell(w)-1} 2^{-i} w[i]$.

int(w) considers w as a rational number in [0, 1] written in binary and computes its value.

int(w) can be written as

$$\operatorname{int}(w) = \sum_{U \subset [\ell(w)]: \operatorname{INIT}_1(U)} \prod_{i \in U} (2^{-1})$$

where INIT₁(U) says that U is an initial segment of $\langle \ell(w), \langle \rangle$ where the last element is in P_1 .

It should be clear that it is very *convenient* and *user friendly* to define word functions as terms in MSOLEVAL_{\mathcal{F}}.

Examples of word functions **NOT**in MSOLEVAL

- (i) The function $sqexp(w) = 2^{\ell(w)^2} = \prod_{(x,y):x=x \land y=y} 2$ is not in MSOLEVAL because the product is over tuples, rather than elements.
- (ii) The function dexp $(w) = 2^{2^{\ell(w)}}$ is not representable in MSOLEVAL_F due to a growth argument.

Characterizing functions defined by Multiplicity Automata

Main Theorem: (N. Labai and J.A.M., 2012)

Let \mathcal{F} be a field, and $f: \Sigma^* \to \mathcal{F}$.

The following are equivalent:

- (i) $f = f_A$ for some Multiplicity Automaton A over \mathcal{F} .
- (ii) $f \in \mathsf{MSOLEVAL}_{\mathcal{F}}$
- (iii) $f \in \mathsf{CMSOLEVAL}_{\mathcal{F}}$
- (iv) $M(\circ, f)$ has finite rank.

Proof: (i) \leftrightarrow (iv) is the Carlyle-Paz Theorem. (ii) \leftrightarrow (iii) follows from CMSOL equals MSOL on words. (iii) \rightarrow (iv) is the Finite Rank Theorem. (i) \rightarrow (ii) is proven using matrix algebra and logic. File:ma-intro Previous attempts of characterizing MA-recognizable functions

Our Main Theorem is an analogue to the Büchi-Elgot-Trakhtenbrot Theorem for multiplicity automata.

There were previous attempts to prove such a theorem using a subset RMSOL of weighted MSOL-formulas rather than MSOL-definable functions.

- M. Droste and P Gastin, Weighted automata and weighted logic, TCS 380 (2007), pp. 69-86.
- M. Droste, W. Kuich and H. Vogler, eds., Handbook of Weighted Automata, Springer 2009

Weighted RMSOL vs. MSOLEVAL

However, there are serious disadvantages in their approach.

- (i) The definition of RMSOL is not a purely syntactic.
- (ii) The formulas are hybrid objects, mixing constants from \mathcal{F} and logical expressions. For instance $\forall x \cdot 2$ is a weighted formula (for 2 = 1 + 1 in a field) which represents the function $2^{\ell(w)}$, and $\forall x \forall y \cdot 2$ is a weighted formula which represents the function $2^{2^{\ell(w)}}$.
- (iii) Seemingly equivalent formulas can represent different functions: $\exists x P_1(x)$ represents the function $\sharp_1(w)$ but $\exists (P(x) \lor P(x))$ represents the function $2 \cdot \sharp_1(w)$.
- (iv) Some of these disadvantages have been corrected in very recent papers byM. Droste and P. Gastin in the Handbook andB. Bollig, P. Gastin , B. Monmege and M. Zeitoun presented at ICALP 2010.

In contrast to these disadvantages, MSOLEVAL_F has the following advantages:

- (i) The expressions are natural and intuitive.
- (ii) The expressions are defined for all formulas of MSOL without any restrictions.
- (iii) If we replace formulas occurring in an expression by equivalent formulas, the word function it represents remains the same.

What else is in the paper?

• If the field \mathcal{F} is replace by a semiring \mathcal{S} a similar result holds.

Instead of the finite rank condition of the Hankel matrix we have to require that the word function is in a finitely generated, stable semimodule.

• We also give a direct translation between RMSOL and MSOLEVAL which uses the syntactic restriction imposed in RMSOL.

More Details and Proofs

If time permits we now discuss the following:

- We discuss the Bilinear Decomposition Theorem for word functions in MSOLEVAL.
- We show how to derive the finite rank of the Hankel matrix using the Bilinear Decomposition Theorem.
- We show how to convert a word function f recognizable by a weighted automaton into an equivalent expression in MSOLEVAL representing f.

The Bilinear Decomposition Theorem

Let $f \in \mathsf{MSOLEVAL}_{\mathcal{F}}$ be a word function $\Sigma^* \to \mathcal{F}$.

We would like to compute $f(u \circ v)$ from f(u) and f(v) only.

Let us discuss two examples with $\Sigma = \{0, 1\}$ and $u, v, w \in \Sigma^*$.

- $\sharp_1(w)$ counts the number of 1's in w and is in MSOLEVAL_F.
- $b_1(w)$ counts the number of blocks of 1's in w. A block of 1's in w is a maximal set of consecutive positions $i \in [\ell(w)]$ in the word w with $P_1(i)$.

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Computing $\sharp_1(u \circ v)$

Clearly we have

 $\sharp_1(u \circ v) = \sharp_1(u) + \sharp_1(v).$

Computing $b_1(u \circ v)$, I

$$b_1(u \circ v) = \begin{cases} b_1(u) + b_1(v) - 1 & P_1(u[\ell(u)]) \text{ and } P_1(v[1]) \\ b_1(u) + b_1(v) & \text{else} \end{cases}$$

To compute $b_1(u \circ v)$ we introduce auxiliary functions from MSOLEVAL_F.

- (i) $f_1(w)$ counts the number of blocks of 1's in w which include the **first** position.
- (ii) $\ell_1(w)$ counts the number of blocks of 1's in w which include the **last** position.
- (iii) $i_1(w)$ counts the number of blocks of 1's in w which exclude the **first and last** position.
- (iv) $fl_1(w)$ counts the number of blocks of 1's in w which contain **both the first and last** position.
- (v) c(w) = 1, the constant function with value 1.

It is easily verified that they are really in $MSOLEVAL_{\mathcal{F}}$.

Computing $b_1(u \circ v)$, II

Clearly, we have

$$b_1(w) = f_1(w) + l_1(w) + i_1(w) + fl_1(w) - fl_1(w) \cdot f_1(w) - fl_1(w) \cdot l_1(w) \quad (1)$$

Furthermore, we have $f_1(w), \ell_1(w) \in \{0, 1\}$ and

$$f_1(u \circ v) = f_1(u) \tag{2}$$

$$\ell_1(u \circ v) = \ell_1(v) \tag{3}$$

$$i_{1}(u \circ v) = i_{1}(u) + i_{1}(v) + \ell_{1}(u) + f_{1}(v) - \ell_{1}(u)f_{1}(v) - fl_{1}(u)fl_{1}(v)$$
(4)
$$fl_{1}(uv) = fl_{1}(u) \cdot fl_{1}(v)$$
(5)

$$l_1(uv) = fl_1(u) \cdot fl_1(v)$$
 (5)

$$c(u \circ v) = 1 \tag{6}$$

Let $B(w) = (f_1(w), \ell_1(w), i_1(w), fl_1(w), c(w)).$

Computing $b_1(u \circ v)$, III

Proposition

There are matrices $M^f, M^l, M^i, M^{fl}, M^c \in \mathcal{F}^{5 \times 5}$ such that

$$f_1(u \circ v) = B(u) \cdot M^f \cdot B(v)^{tr}$$
(7)

$$\ell_1(u \circ v) = B(u) \cdot M^l \cdot B(v)^{tr}$$
(8)

$$i_1(u \circ v) = B(u) \cdot M^i \cdot B(v)^{tr}$$
(9)

$$fl_1(u \circ v) = B(u) \cdot M^{fl} \cdot B(v)^{tr}$$
(10)

$$c(u \circ v) = B(u) \cdot M^c \cdot B(v)^{tr}$$
(11)

Computing $b_1(u \circ v)$, IV

Using the equations (1) - (6) one easily verifies that

Conclusion

- (i) The Hankel matrix $H(\sharp_1)$ of \sharp_1 has rank 2 due to equation ()
- (ii) The Hankel matrices $H(b_1), H(f_1), H(\ell_1)$ and $H(i_1)$ of b_1, f_1, ℓ_1, i_1 have rank at most 5.

Bilinear Decomposition Theorem (BDT) for Word Functions

The two examples are typical in the following sense:

Theorem:

Let $f \in MSOLEVAL_{\mathcal{F}}$ be a word function $\Sigma^* \to \mathcal{F}$ of quantifier rank q(f). There is a finite sequence $F = (g_1, \ldots, g_{\alpha(f)})$ of functions in $MSOLEVAL_{\mathcal{F}}$ of size $\alpha(f)$ and for each g_i there is a matrix $M^{(i)} \in \mathcal{F}^{q(f) \times q(f)}$ such that

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(i) f \in F and
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(ii) g_i(u \circ v) = F(u) \cdot M^{(i)}F(v)^{tr}.
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 $\alpha(f)$ actually only depends on q(f) but grows very quickly.

The full proof is in B. Courcelle, J.A.M. and U. Rotics, DAM 2001.

The bilinear version was only formulated later in J.A.M., Annals of Pure and Applied Logic, 2004, but uses the same proof.

Here we merely note that F can be chosen to consist of all the functions in MSOLEVAL_F of quantifier rank at most q(f).

This is a rough estimate. The examples above show that F can often be much smaller.

Finite Rank Theorem for Word Functions

Using the BDT we get

Theorem (B. Godlin, T. Kotek and J.A.M., 2008)

Let $f \in MSOLEVAL_{\mathcal{F}}$ be a word function $f : \Sigma^* \to \mathcal{F}$ with all the formulas of quantifier rank at most q(f).

Then the Hankel matrix H(f) has rank at most $\alpha(f)$.

Proof (of the Main Theorem), I

Let A be a weighted automaton of size r over \mathcal{F} for words in Σ^* given by

- (i) Two vectors $\alpha, \gamma \in \mathcal{F}^r$, and
- (ii) for each $\sigma \in \Sigma$ a matrix $\mu_{\sigma} \in \mathcal{F}^{r \times r}$.

For a word $w = \sigma_1 \sigma_2 \dots \sigma_{\ell(w)}$ the automaton A defines the function

$$f_A(w) = \alpha^T \mu_{\sigma_1} \cdot \ldots \cdot \mu_{\sigma_{\ell(w)}} \cdot \gamma.$$
(15)

We have to show that $f_A \in MSOLEVAL_F$.

To unify notation we define

$$M(i, j, \sigma) = (\mu_{\sigma})_{i,j}.$$
(16)

Furthermore, the word w is given as a function $w : [\ell(w)] \to \Sigma$.

Proof, II

Using Equation 15 and matrix algebra we get

$$f_{A}(w) = \sum_{\pi:[n+2]\to[r]} \alpha_{\pi(1)} \cdot [M(\pi(1), \pi(2), w(1)) \cdot \ldots \cdot M(\pi(n), \pi(n+1), w(n))] \cdot \gamma_{\pi(n+2)} = \sum_{\pi:[n+2]\to[r]} \alpha_{\pi(1)} \cdot \left(\prod_{v\in[n]} M(\pi(v), \pi(v+1), w(v))\right) \cdot \gamma_{\pi(n+2)}$$
(17)

Proof, III

To convert Equation (17) into an expression in MSOLEVAL(Σ) we use the following lemmas:

Let S be any set and $\pi : S \to [r]$ be a function. π induces a partition of S into sets $U_1^{\pi}, \ldots, U_r^{\pi}$ by $U_i^{\pi} = \{s \in S; \pi(s) = i\}$. Conversely, every partition $\mathcal{U} = (U_1, \ldots, U_r)$ of S induces a function $\pi_{\mathcal{U}}$ by setting $\pi_{\mathcal{U}}(s) = i$ for $s \in U_i$.

Lemma 1

Let $M(\pi)$ be any function depending on pi.

$$\sum_{\pi:S\to[r]} M(\pi) = \sum_{\mathcal{U}} M(\pi_{\mathcal{U}}) = \sum_{U_1,\dots,U_r:Partition(U_1,\dots,U_r)} M(\pi_{\mathcal{U}})$$
(18)

where \mathcal{U} ranges over all partition of S into r sets $U_i : i \in [r]$.

Clearly, $Partition(U_1, \ldots, U_r)$ can be written in MSOL.

Proof, IV

To convert the factors $\alpha_{\pi(1)}$ and $\gamma_{\pi(n+2)}$ we proceed as follows:

Lemma 2

Let α_i be the unique value of the coordinate of α such that $1 \in U_i$. Similarly, let γ_i be the unique value of the coordinate of γ such that $n + 2 \in U_i$.

$$\alpha_{\pi(1)} = \prod_{i=1}^{r} \prod_{1 \in U_i} \alpha_i$$

$$\gamma_{\pi(n+2)} = \prod_{i=1}^{r} \prod_{n+2 \in U_i} \gamma_i$$
(19)
(20)

Proof:

First we note that, as \mathcal{U} is the partition induced by π , the restriction of π to U_i is constant for all $i \in [r]$. Next we note that the product ranging over the empty set gives the value 1. Q.E.D.

Proof, V

Similarly, to convert the factor $\prod_{v \in [n]} M(\pi(v), \pi(v+1), w(v))$ use following lemma:

Lemma 3

Let $M_{i,j,w(v)}$ be the unique value of the (i, j)-entry of the matrix $\mu_{w(v)}$ such that $v \in U_i$ and $v + 1 \in U_j$.

$$\prod_{v \in [n]} M(\pi(v), \pi(v+1), w(v)) = \prod_{i,j=1}^{r} \prod_{v \in U_i, v+1 \in U_j} M_{i,j,w(v)}$$
(21)

By writing $U_i(v)$ instead of $v \in U_i$ it is not difficult to see that the monomials of the Lemmas 1, 2 and 3 are indeed in MSOLEVAL_F.

Using that $MSOLEVAL_{\mathcal{F}}$ is closed under products and using Lemmas 1, 2 and 3 we complete the proof