2. Describing Constraints by Automata

"Besides, that's not a regular rule: you invented it just now."

Alice's Adventures in Wonderland
LEWIS CARROLL

This chapter recapitulates the standard theory of automata (see also, e.g., [36]). We introduce the reader to finite automata and regular languages (Section 2.1) and then we define the AUTOMATON constraint predicate in three stages: first its particular case that is also known as the REGULAR constraint predicate [43] (Section 2.2), and then two orthogonal extensions, namely predicate automata (Section 2.3) and automata with accumulators (Section 2.4). Finally, we compose the two extensions into predicate automata with accumulators (Section 2.5).

2.1 Finite Automata and Regular Languages

A deterministic finite automaton (DFA) [36], or automaton for short, is a tuple $\langle Q, \Gamma, \delta, \rho_0, Q_a \rangle$ where Q is the finite set of states; Γ is the finite alphabet; ρ_0 is a state in Q denoting the initial state; Q_a is a subset of Q denoting the accepting states; and δ is a total function from $Q \times \Gamma$ to Q denoting the transition function. If $\delta(\rho, a) = \rho'$, then we say that there is a transition from state ρ to state ρ' that consumes alphabet symbol a; this is here often written as:

$$\rho \xrightarrow{a} \rho'$$

A *word* is here a sequence of symbols from a given alphabet. Let Γ^* denote the infinite set of words built from Γ , including the empty word, denoted ε . The *extended transition function* $\widehat{\delta}: Q \times \Gamma^* \to Q$ for words (instead of symbols) is recursively defined by $\widehat{\delta}(\rho, \varepsilon) = \rho$ and $\widehat{\delta}(\rho, wa) = \delta(\widehat{\delta}(\rho, w), a)$ for a word w and symbol a. Note that both δ and $\widehat{\delta}$ are total functions. A word $w = a_1 a_2 \cdots a_{n-1} a_n$ is *accepted* by the automaton if there is a chain of transitions:

$$\rho_0 \xrightarrow{a_1} \rho_1 \xrightarrow{a_2} \dots \xrightarrow{a_{n-1}} \rho_{n-1} \xrightarrow{a_n} \rho_n$$

¹Automata with accumulators are called counter automata in Paper II, and memory-DFAs in Paper III and Paper V.

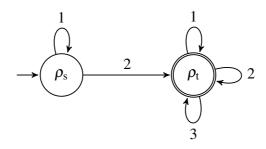


Figure 2.1. DFA for the regular expression 1*2(1|2|3)*.

such that $\rho_n \in Q_a$, that is if $\widehat{\delta}(\rho_0, w) \in Q_a$.

One often uses pictures to define finite automata. For example, in Figure 2.1, we define an automaton with two states, $Q = \{\rho_s, \rho_t\}$, represented by circles, and an alphabet of three symbols, $\Gamma = \{1,2,3\}$, on the transitions. The initial state $\rho_0 = \rho_s$ is indicated by an arrow coming from nowhere, and an accepting state is represented by a double circle, and so $Q_a = \{\rho_t\}$. The transition function is represented by the annotated arrows, that is $\delta(\rho, a) = \rho'$ if there is an arrow from ρ to ρ' annotated with a. For each state, there is one outgoing arrow per alphabet symbol; any missing arrow is assumed to go to an implicit non-accepting state, on which there is a self-looping arrow for every symbol of the alphabet, so that no accepting state is reachable from that state. For example, in Figure 2.1, the missing transition from state ρ_s on symbol 3 goes to such an implicit non-accepting state.

A language is, in the formal sense, a set of words together with a set of formation rules. A regular language is a language that can be defined using a regular expression. Regular expressions describe patterns over words; for example, the regular expression $1^*2(1|2|3)^*$ over the alphabet $\Gamma = \{1,2,3\}$ defines the set of words that start with zero or more 1s, followed by exactly one 2, and ending with any number of symbols, possibly zero, from Γ . We say that $1^*2(1|2|3)^*$ defines a regular language. We denote the language defined by a regular expression σ by $\mathcal{L}(\sigma)$. For example, the words 2 and 121 are words in $\mathcal{L}(1^*2(1|2|3)^*)$, whereas the words 11 and 13 are not. We can also relate regular languages to automata: a language is regular if and only if every word in the language is accepted by a deterministic finite automaton. For this reason, we say that an automaton accepts a regular language \mathcal{L} , since it accepts all the words in \mathcal{L} and rejects all the other ones. For example, the automaton in Figure 2.1 accepts the language of the regular expression $1^*2(1|2|3)^*$.

A deterministic finite transducer [48] is a tuple $\langle Q, \Gamma, \Gamma', \delta, \rho_0, Q_a \rangle$, where Q is the finite set of states, Γ is the finite input alphabet, Γ' is the finite output alphabet, $\delta: Q \times \Gamma \to Q \times \Gamma'^*$ is the transition function, which must be total, $\rho_0 \in Q$ is the initial state, and $Q_a \subseteq Q$ is the set of accepting states. When $\delta(\rho, a) = \langle \rho', a' \rangle$, there is a transition from state ρ to state ρ' upon consuming the input symbol a and producing the sequence a' of output symbols: we write

this as $\rho \xrightarrow{a:a'} \rho'$. Note that a deterministic finite automaton is a transducer without an output alphabet. In a graphical representation of a transducer, a transition is depicted by an arrow between two states, possibly the same, and is annotated by a consumed input symbol, followed by a colon and a sequence of produced output symbols (see Figure 3.4 for an example).

2.2 Describing Constraints by Deterministic Finite Automata

Any constraint (on a sequence of decision variables) whose extensional definition forms a regular language can be described by an automaton. In fact, any constraint on a finite sequence of decision variables that range over finite domains can be described by an automaton, since every finite language is a regular language. The REGULAR(\mathcal{A}, \mathcal{V}) constraint [13, 43] holds if the constraint described by the deterministic finite automaton \mathcal{A} (or its equivalent regular expression) holds for the sequence \mathcal{V} of decision variables, that is if \mathcal{A} accepts the sequence of values of \mathcal{V} .

In practice, an automaton may however have a number of states that is exponential in the number of decision variables of the constraint, such as for the ALLDIFFERENT constraint predicate, as discussed in [43].

A REGULAR(\mathcal{A}, \mathcal{V}) constraint can be implemented either via a specialised propagator [43] or via decomposition into a conjunction of constraints [13]. We here take the latter approach because it will be more convenient when defining the extensions in Sections 2.3 and 2.4. For a given automaton $\mathcal{A} = \langle Q, \Gamma, \delta, \rho_0, Q_a \rangle$, we define a new constraint predicate T extensionally by the following set:

$$\{\langle q, a, q' \rangle \mid q \xrightarrow{a} q'\} \tag{2.1}$$

That is, T(q, a, q') is satisfied whenever there is a transition in A from state q to state q' that consumes symbol a. A REGULAR $(A, \langle v_1, \ldots, v_n \rangle)$ constraint is then decomposed into the following conjunction of n+2 constraints, called the *transition constraints*:

$$q_0 = \rho_0 \wedge \mathrm{T}(q_0, v_1, q_1) \wedge \cdots \wedge \mathrm{T}(q_{n-1}, v_n, q_n) \wedge q_n \in Q_a$$
 (2.2)

where $q_0, q_1, \ldots, q_{n-1}, q_n$ are new decision variables, called the *state variables*, with domain Q. For contrast, we call v_1, \ldots, v_n the *problem variables*.

This decomposition actually works unchanged for *non-deterministic finite* automata (NFA), where δ is a relation rather than a total function (for example, see Figure 2.2), but we have elected to restrict our focus to deterministic ones, in order to ease the notation.

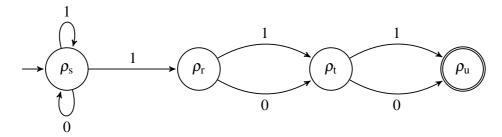


Figure 2.2. NFA for the regular expression $(0|1)^*1(0|1)^2$: all 0/1 sequences that have a 1 two characters from the end of the sequence.

2.3 Describing Constraints by Predicate Automata

The automata in [13] are more powerful than those in [43]: The alphabet symbols can be predicates on variables, and all predicates on an accepting path must be satisfied.

The definition presented here is parametrised by a suitable set of predicates. Let \mathbf{Pred}_k be a set of k-ary predicates in some suitable language. That is, a predicate takes a vector, \mathcal{P} , of k values.

A *k-ary predicate automaton* is a tuple $\langle Q, \Gamma, \delta, \phi, \rho_0, Q_a \rangle$, where $Q, \Gamma, \delta, \rho_0$, and Q_a are exactly as for a deterministic finite automaton, and ϕ is a function from Γ to \mathbf{Pred}_k . For all *k*-ary value vectors \mathcal{P} and all distinct symbols a_1 and a_2 of Γ , we must have that $\phi(a_1)(\mathcal{P}) \wedge \phi(a_2)(\mathcal{P})$ is false (that is, any two predicates must be mutually exclusive). A sequence of *k*-ary vectors of values $\mathcal{P}_1\mathcal{P}_2\cdots\mathcal{P}_{n-1}\mathcal{P}_n$ is *accepted* by the automaton if there exists a chain of transitions

$$\rho_0 \xrightarrow{a_1} \rho_1 \xrightarrow{a_2} \dots \xrightarrow{a_{n-1}} \rho_{n-1} \xrightarrow{a_n} \rho_n$$

such that $\rho_n \in Q_a$ and $\phi(a_i)(\mathcal{P}_i)$ is true for all $1 \leq i \leq n$. Such a chain of transitions can be written as

$$\rho_0 \xrightarrow{\phi(a_1)(\mathcal{P}_1)} \rho_1 \xrightarrow{\phi(a_2)(\mathcal{P}_2)} \dots \xrightarrow{\phi(a_{n-1})(\mathcal{P}_{n-1})} \rho_{n-1} \xrightarrow{\phi(a_n)(\mathcal{P}_n)} \rho_n$$

Again, we often define k-ary predicate automata by pictures. The convention is similar to normal finite automata, except that the transition labels are predicates. We assume that each distinct predicate is associated with a distinct symbol of the alphabet Γ , and that the function ϕ is defined by the predicate labels in the picture.

For example, in Figure 2.3, the function ϕ could be defined by lambda expressions as follows: $\phi(1) = \lambda x, y : x = y, \ \phi(2) = \lambda x, y : x < y, \ \text{and} \ \phi(3) = \lambda x, y : x > y$. Consider the constraint that the sequence of decision variables \mathcal{V} be lexicographically less than the sequence of decision variables \mathcal{W} , which is denoted by $\mathcal{V} <_{\text{lex}} \mathcal{W}$. For the fixed sequences $\mathcal{V} = \langle 1, 2, 5, 6 \rangle$ and $\mathcal{W} = \langle 1, 3, 4, 7 \rangle$, the sequence $\langle 1, 1 \rangle \langle 2, 3 \rangle \langle 5, 4 \rangle \langle 6, 7 \rangle$ of binary vectors, obtained by zipping \mathcal{V} and \mathcal{W} together, is accepted by the binary predicate automaton

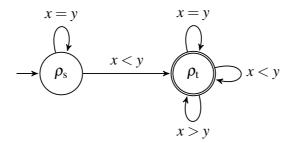


Figure 2.3. A k-ary predicate automaton with k = 2 describing the $<_{lex}$ constraint predicate.

(k = 2) in Figure 2.3 because the transition chain

$$\rho_{\rm s} \xrightarrow{1=1} \rho_{\rm s} \xrightarrow{2<3} \rho_{\rm t} \xrightarrow{5>4} \rho_{\rm t} \xrightarrow{6<7} \rho_{\rm t}$$

ends in the accepting state ρ_t .

Given a predicate automaton $\langle Q, \Gamma, \delta, \phi, \rho_0, Q_a \rangle$, the automaton $\langle Q, \Gamma, \delta, \rho_0, Q_a \rangle$ is referred to as the *underlying automaton* of the predicate automaton. For example, the automaton in Figure 2.1 is the underlying automaton of the predicate automaton in Figure 2.3.

In [13], the AUTOMATON(\mathcal{A}, \mathcal{V}) constraint holds if and only if the constraint described by the automaton \mathcal{A} holds for the sequence \mathcal{V} of decision variables, where \mathcal{A} is a predicate automaton implemented with the help of reification. The constraint predicate T defined in (2.1) is used for the following n+2 transition constraints:

$$q_0 = \rho_0 \wedge \mathsf{T}(q_0, S_1, q_1) \wedge \dots \wedge \mathsf{T}(q_{n-1}, S_n, q_n) \wedge q_n \in Q_a$$
 (2.3)

These transition constraints are like (2.2), but are expressed for *new* decision variables $S_1, ..., S_n$, which are connected as follows to the sequence of problem variables \mathcal{V} via the automaton predicates and reification: given an n-length sequence $\mathcal{V} = \langle \mathcal{V}_1, ..., \mathcal{V}_n \rangle$ of k-ary vectors of problem variables, we add the following n constraints, called the *signature constraints*:

$$\bigwedge_{i=1}^{n} \left(\bigwedge_{a \in \Gamma} \left(S_i = a \Leftrightarrow \phi(a)(\mathcal{V}_i) \right) \right)$$
 (2.4)

where the S_i are called the *signature variables*, with domain Γ . Hence \mathbf{Pred}_k contains whatever can be implemented as reified constraints in the underlying CP solver (note that most global constraint predicates can be reified [12]). For example, in Figure 2.3, the binary predicate automaton on the two sequences of variables $\mathcal{V} = \langle v_1, \dots, v_n \rangle$ and $\mathcal{W} = \langle w_1, \dots, w_n \rangle$ requires the transition constraints (2.3) and the following signature constraints for all $1 \le i \le n$:

$$(S_i = 1 \Leftrightarrow v_i = w_i) \land (S_i = 2 \Leftrightarrow v_i < w_i) \land (S_i = 3 \Leftrightarrow v_i > w_i)$$

2.4 Describing Constraints by Automata with Accumulators

While the class of constraint predicates that can be described by (predicate) automata is large (60 of the 381 constraint predicates of the *Global Constraint Catalogue* [10] are described that way), it is often the case that (predicate) automata are very large or specific to a problem instance. The second extension in [13] is the use of integer accumulators² that are initialised at the start and evolve through accumulator-updating operations coupled to the transitions of the automaton. Such automata with accumulators allow the capture of non-regular languages and yield (even for regular languages) automata that are often much smaller if not instance-independent and enable constraint predicates to be described succinctly or generically. The two extensions are orthogonal and can be composed, so we define this second extension in isolation.

Again, we give a definition that is parametric, namely on the class of accumulator-updating functions. An accumulator-updating operation consists of a sequence of assignments to some accumulators (the accumulators without assignments are left unchanged), possibly guarded by a condition on the current accumulator values and the variables. Let $\mathbf{AccUpdate}_{\ell}$ be a set of ℓ -ary accumulator-updating functions. That is, given a function $\mathbf{\psi} \in \mathbf{AccUpdate}_{\ell}$ and a vector of accumulators $\mathcal{C} \in \mathbb{Z}^{\ell}$, we have that $\mathbf{\psi}(\mathcal{C})$ is a new vector in \mathbb{Z}^{ℓ} .

An ℓ -ary automaton with accumulators is a tuple $\langle Q, \Gamma, \delta, \rho_0, \mathcal{C}_0, Q_a, \alpha \rangle$ where Q, Γ , ρ_0 , and Q_a are exactly as for a deterministic finite automaton; vector \mathcal{C}_0 has the initial values of a vector \mathcal{C} of ℓ accumulators; and δ is a function from $Q \times \Gamma$ to $Q \times \mathbf{AccUpdate}_{\ell}$. If $\delta(\rho, a) = (\rho', \psi)$ and $\psi(\mathcal{C}) = \mathcal{C}'$, then we write

$$(\rho, \mathcal{C}) \xrightarrow{a} (\rho', \mathcal{C}')$$

and similarly for its extended version $\hat{\delta}$. A word $a_1 a_2 \cdots a_{n-1} a_n$ is accepted by the automaton if there is a chain of transitions

$$(\rho_0, \mathcal{C}_0) \xrightarrow{a_1} (\rho_1, \mathcal{C}_1) \xrightarrow{a_2} \dots \xrightarrow{a_{n-1}} (\rho_{n-1}, \mathcal{C}_{n-1}) \xrightarrow{a_n} (\rho_n, \mathcal{C}_n)$$

such that $\rho_n \in Q_a$. Finally, $\alpha \colon Q_a \times \mathbb{Z}^k \to \mathbb{Z}$ is called the *acceptance function* and transforms the accumulators at an accepting state into an integer. Given a word w, the automaton with accumulators returns $\alpha(\widehat{\delta}(\langle \rho_0, \mathcal{C}_0 \rangle, w))$ if w is accepted. Note that δ , $\widehat{\delta}$, and α are total functions.

As with automata, one often uses pictures to define automata with accumulators. The set Q of states, the set Q_a of accepting states and the initial state ρ_0 are defined exactly as for an automaton. The transition function is also defined by the annotated arrows, but the label on the arrow of a transition consists of a symbol followed by an accumulator-updating operation between curly braces. That is $\delta(\rho, a) = (\rho', \psi)$ if there is an arrow from ρ to ρ' annotated with $a \{ \psi \}$.

²Accumulators are called counters in [13] and in Paper II.

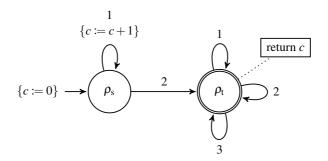


Figure 2.4. Automaton with $\ell = 1$ accumulator for the regular expression $1^*2(1|2|3)^*$. Accumulator c maintains the length of the longest prefix matching the regular expression 1^* of the sequence of symbols consumed so far.

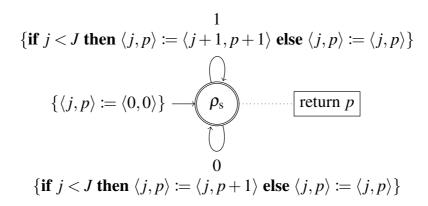


Figure 2.5. An ℓ -ary automaton with accumulators with $\ell=2$ accumulators describing the JTHNONZEROPOS (\mathcal{V},J,P) constraint [10], which holds if and only if P is the position (counting from 1) of the J^{th} non-zero element of the sequence $\mathcal{V}=\langle V_1,\ldots,V_n\rangle$. Accumulator j maintains the number of non-zero values among the J first non-zero elements of \mathcal{V} , while accumulator p maintains the number of all values within that prefix of \mathcal{V} . Upon acceptance, the final value of the vector of accumulators $\langle j,p\rangle$ must be $\langle J,P\rangle$. The signature constraints are $S_i=0\Leftrightarrow V_i=0$ and $S_i=1\Leftrightarrow V_i\neq 0$.

For example, in Figure 2.4, the self-loop on ρ_s depicts that $\delta(\rho_s,1)=(\rho_s,\langle c+1\rangle)$ for all c. If an update corresponds to the identity function, then we do not depict it; for example, the three self-loops on ρ_t have no depicted updates, as $\langle c \rangle \coloneqq \langle c \rangle$. If an update involves only one accumulator, then we omit the angled brackets; for example, the self-loop on ρ_s has $c\coloneqq c+1$ instead of $\langle c \rangle \coloneqq \langle c+1 \rangle$. The acceptance function α transforms the vector of accumulators $\langle c \rangle$ at ρ_t into c, and is depicted by a box linked to ρ_t by a dotted line. Note that an accumulator-updating operation can also be guarded by a condition on the current accumulator values and the problem variables, as can be seen in Figure 2.5.

In [13], constraint predicates described by automata with accumulators are decomposed into transition constraints that are slightly extended to include information about the values of the accumulators. We define the transition

constraint predicate T extensionally by the following set:

$$\{\langle q, \mathcal{C}, a, q', \mathcal{C}' \rangle \mid (q, \mathcal{C}) \xrightarrow{a} (q', \mathcal{C}')\}$$

An AUTOMATON($\mathcal{A}, \mathcal{V}, R$) constraint on a sequence of n problem variables, with $\mathcal{V} = \langle v_1, \dots, v_n \rangle$, and an result parameter (either an integer constant or a decision variable), R, is then decomposed into the following conjunction of n+4 transition constraints:

$$q_0 = \rho_0 \wedge c_0 = \mathcal{C}_0 \wedge \mathrm{T}(q_0, c_0, v_1, q_1, c_1) \wedge \cdots \\ \wedge \mathrm{T}(q_{n-1}, c_{n-1}, v_n, q_n, c_n) \wedge q_n \in Q_{\mathbf{a}} \wedge \alpha(c_n) = R$$
 (2.5)

where q_0, \ldots, q_n are state variables, with domain Q, while c_0, \ldots, c_n are vectors of new integer decision variables, called *accumulator variables*.

Upon acceptance, we must have $\alpha(c_n) = R$; initially, we have $c_0 = C_0$ where C_0 is a parameter of the automaton. It is also important not to mix up the vectors of variables c_0, \ldots, c_n with the vector c of accumulators of the automaton.

By abuse of language, when there is $\ell = 1$ accumulator, we often refer to vector C_0 as the initial value (rather than the vector with the initial value), to vector C as an accumulator value, and to vector c_i as an accumulator variable.

2.5 Describing Constraints by Predicate Automata with Accumulators

A $\langle k,\ell \rangle$ -ary predicate automaton with accumulators, or simply automaton, is an automaton that is both a k-ary predicate automaton and an ℓ -ary automaton with accumulators. A $\langle k,\ell \rangle$ -ary predicate automaton with accumulators is a tuple $\langle Q,\Gamma,\delta,\phi,\mathcal{C}_0,\rho_0,Q_a,\alpha \rangle$ where Q,Γ,ρ_0 , and Q_a are exactly as for a automaton; ϕ is a function from Γ to \mathbf{Pred}_k ; vector \mathcal{C}_0 has the initial values of the ℓ accumulators; and δ is a function from $Q \times \Gamma$ to $Q \times \mathbf{AccUpdate}_{\ell}$.

For example, in Figure 2.6, we define a predicate automaton with accumulators where $Q = \{\rho_s, \rho_t\}$ has two states, $\Gamma = \{1, 2, 3\}$ is an alphabet of three symbols, ϕ is the function defined by $\phi(1) = \lambda x, y : x = y$, $\phi(2) = \lambda x, y : x < y$, and $\phi(3) = \lambda x, y : x > y$, the accumulator c has the initial value $C_0 = \langle 0 \rangle$, $Q_a = \{\rho_t\}$ has one accepting state, and the transition function δ is as indicated with the annotated arrows. The arrow indicating the initial state of the automaton is preceded by the sequence of initialising assignments of the accumulators. The label on the arrow of a transition consists of a predicate followed by an accumulator-updating operation between curly braces.

Since a predicate automaton with accumulators consumes the signature variables S_i instead of the k-ary vectors of problem variables \mathcal{V}_i , the transition constraints (2.5) given in Section 2.4 for an AUTOMATON($\mathcal{A}, \mathcal{V}, R$) constraint,

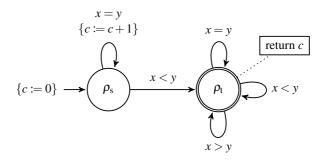


Figure 2.6. A $\langle 2,1\rangle$ -ary predicate automaton with accumulators describing a constraint predicate on two sequences of decision variables $\mathcal V$ and $\mathcal W$ which holds if and only if $\mathcal V<_{\rm lex}\mathcal W$ holds and accumulator c denotes the length of the longest common prefix between $\mathcal V$ and $\mathcal W$.

with $\mathcal{V} = \langle \mathcal{V}_1, \dots, \mathcal{V}_n \rangle$, are transformed into the following:

$$q_0 = \rho_0 \wedge c_0 = \mathcal{C}_0 \wedge \operatorname{T}(q_0, c_0, S_1, q_1, c_1) \wedge \cdots \\ \wedge \operatorname{T}(q_{n-1}, c_{n-1}, S_n, q_n, c_n) \wedge q_n \in Q_a \wedge \alpha(c_n) = R \quad (2.6)$$

Even though the transition constraints are defined extensionally, they can be efficiently implemented using the CASE constraint predicate of SICStus Prolog [26] and the ELEMENT constraint predicate: see [9] for details.

We collectively refer to the signature variables S_i , accumulator variables c_i , and state variables q_i as the *induced variables* of the automaton.

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