Centre for Data Analytics



A Transducer-Based Model for Representing Functional Constraints on Integer Sequences

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Key Ideas

- Describe mechanism to define many global constraints declaratively
- Seen as functional constraints on a sequence $c(R, [X_1, X_2, ..., X_n])$
- Automatically generate constraints from description
- Based on automata with registers
- Provides basis for much more (not covered here):
 - Bounds
 - Redundant constraints
 - Interaction of constraints

Examples

- nb_strictly_decreasing_sequence(2,[1,1,0,0,1,0,0,1])
- max_width_group(2,[0,1,0,1,1])
- $nb_stretch(6, [0, 1, 1, 1, 0, 1, 0, 1])$

Why?

- Constraint solvers depend on global constraints for performance
- But each constraint is an algorithm on its own
- Too many to implement
- Impossible to be confident in correctness
- Too many to remember
- We need a more systematic way to describe and implement global constraints

Note

- We describe the process based on integer sequences
- The derived mechanism also works as a constraint on finite domain variables

Computing with Integer Sequences



Insight Centre for Data Analytics

Signature Definition

- Time-Series: binary <, =, >
- Unary example: \in, \notin for value set *V*, max_width_group
- Other binary example : =, \neq , nb_stretch
- k-ary: Compute signature value from sub-sequences of length *k*
- Unary classification
- Extended binary $\ll, <, =, >, \gg$

The Scheme



But you did all this already for time-series?

- This is a second-generation study
- Time-series were a special case
 - Specific, binary signature (<,=,>)
 - Initially hand-crafted transducers
 - Feature/Aggregator set limited
 - Only one feature considered per constraint
- Some issues were not dealt with elegantly
 - Mix-up of pattern recognized and feature value computed
 - Some pattern were not expressible

Compare: Old vs New

<code>bump_on_decreasing_sequence</code>, Reg. expr '>><>>', Signature '<=>'



Example Recognition '>><>>'



The Generated Automaton

- Based on the generated transducer
- Five registers
 - Level 4: aggregated value h
 - Level 3: aggregated value g
 - Level 2: feature value of current confirmed, but uncompleted pattern
 - Level 1: potential feature value of unconfirmed pattern
 - Level 0: potential feature value of unconfirmed pattern

The Micro-Instructions

micro instruction register updates

$\texttt{compute}(\ell, b, v)$: if $b = 0$ then $V_{\ell} \leftarrow \phi_{\ell}(V_{\ell}, v)$ else $V_{\ell} \leftarrow \phi_{\ell}(V_{\ell}, -v)$	
$\texttt{reset}(\ell)$: for $k \in [0, \ell]$ do $V_k \leftarrow \mathrm{id}_k$	
$\texttt{transmit}(c, b, \ell)$: if $c = 1$ then $V_{\ell+1} \leftarrow V_{\ell}$	
$\mathtt{set}(\ell,k)$	else if $b = 1$ then $V_{\ell+1} \leftarrow \phi_{\ell+2}$	$(V_{\ell+1}, V_\ell)$
	else $V_{\ell+1} \leftarrow \phi_{\ell+1}$	$(V_{\ell+1}, V_{\ell})$
	: if $before + 1 - k > 0$ then	$V_{\ell} \leftarrow \mathrm{id}_{\ell}$
	else if $before + 1 - k = 0$ then	$V_{\ell} \leftarrow \delta^i_f$
	else	$V_{\ell} \leftarrow \phi_{\ell}(\delta_f^{i-k+1+before}, \dots, \delta_f^i)$

The Macro-Instructions

letter	precondition	macro instruction code
$\mathbf{maybe}_{\mathrm{b}}$	$: \begin{pmatrix} s \notin skip \land \\ d > before \end{pmatrix}$	$\texttt{compute}\left(1,0,\delta_{f}^{i}\right),\texttt{transmit}(0,0,0),\texttt{reset}(0)$
	$\begin{pmatrix} s \in skip \land \\ d > before \end{pmatrix}$	$\texttt{compute}\left(0,0,\delta_{f}^{i}\right)$
$\mathbf{maybe}_{\mathrm{r}}^{k}$:	reset(1), set(1,k)
$\mathbf{out}_{\mathrm{r}}$:	reset(1)
found		$\texttt{compute}(1, \textit{balance}, \kappa), \texttt{transmit}(1, 0, 1), \texttt{reset}(1)$
$\mathbf{maybe}_{\mathrm{a}}$:	$\texttt{compute}\left(1, \textit{balance}, \delta_{f}^{i+a-1-after} ight)$
in	:	compute $(1, balance, \delta_f^{i+a-1-after})$, transmit $(0, 0, 1)$, reset (1)
\mathbf{end}	:	transmit(0, balance, 2), transmit(0, 0, 3), reset(2)

Limitations

- Complexity of scheme
 - Initial learning curve
 - Amortized over many constraints
- No guarantees about consistency levels achieved
 - More reasoning possible (similar to time-series results)
 - Better correct and slow than fast and wrong