Decomposition Structures for Soft Constraints Evaluation Problems

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Outline

- Introduction of an algebraic framework for representing and solving Soft Constraint Evaluation Problems
 - Syntax inspired by process algebras
- Correspondence between tree decompositions and terms of the SCEP algebra
- Polyadic soft constraints as an interpretation of the given algebra
- Examples of application

Constraint Satisfaction Problems and dynamic programming

- Constraint Satisfaction Problems can be represented as (hyper)graphs
 - Intuitively, given a domain of values, a set of variables and a set of constraints, we want to find an assignment of the variables s.t. all the constraints are satisfied
 - Further operations required to combine or remove constraints
 - CSP solving is NP-complete
- Dynamic programming applies well to CSPs:
 - We can decompose the problem and then repeatedly solve the subproblems
 - Deciding the best decomposition is known as secondary optimization problem: NP-complete

Soft constraints

- More flexible than classical constraints
 - Each soft constraint associates a certain assignement to a value in a poset
 - ► To solve an SCSP we seek for an assignment of the variable s.t. the total amount is minimized or maximized, due to the fact that we express negative or positive preferences
- SCSPs are difficult to represent and compose
 - Need for a generalisation which allows for representing both the problem's structure and the solution process via an algebraic framework
 - Decomposition and solving are thus correct by construction

Algebraic structures for soft constraints

- lacktriangle Absortive semirings (also known as c-semirings) $\langle S, \oplus, \otimes, 0, 1 \rangle$
 - ▶ The ⊕ operator is idempotent and 1 is its absorbing element
 - ► The ⊗ operator is associative and commutative
- lackbox \oplus is often used to induce the order on the elements of such a semiring
 - ▶ We have $a \le b \iff a \oplus b = b$
- ► We use ⊗ to combine constraints
- ▶ We may additionaly require some kind of "subtraction"
 - Based on residuation theory
 - ► The residuation operator is a kind of weak inverse of ⊗

Algebraic structures for soft constraints

- Properties of semirings are useful for manipulating soft constraints
 - Associativity and commutativity of ⊗ guarantee the order of combination to be irrelevant
 - Absortivity enforces the fact that adding constraints decreases the number of solutions
 - ► The existence of the 0 element represents the crisp feature of total dislike of any solution involving a certain assignement
 - The unit 1 allows for modelling the crisp feature of "indifference" for a certain constraint

Polyadic Soft Constraints

- ▶ A formalism to manipulate soft constraints based on polyadic algebras
 - Two families of operators, called cylindrification and polyadic substitution model variables hiding and parameters passing
 - ► Allows for a polynomial representation of soft constraints
- ► This algebra respects both the weak and the strong SCEP specification, thus representing an interpretation of the initial algebra

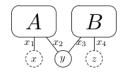
Soft Constraints and concrete networks

- ▶ We assume \mathcal{E}_C to be a ranked alphabet of soft constraints, equipped with an arity function $ar: \mathcal{E}_C \to \mathbb{N}$ and a set of variables \mathbb{V}
- ► A concrete network is a pair I ► N where
 - $N = (V_N, E_N, a_N, lab_N)$ is a labelled hypergraph over \mathcal{E}_C with no isolated vertices
 - ► Vertices are a subset of V
 - I is a set of interface variables
- ▶ We also assume that \mathcal{E}_C has a function $var : \mathcal{E}_C \to \mathbb{V}^*$ which assigns a tuple of distinct canonical variables to each constraint, i.e. $var(A) \cap var(B) = \emptyset$ if $A \neq B$

Example

- A, B constraints with ar(A) = ar(B) = 2
- $var(A) = (x_1, x_2),$ $var(B) = (x_3, x_4)$
- $V_N = \{x, y, z\},\ E_N = \{e_1, e_2\}$
- $a_N(e_1) = \langle x, y \rangle,$ $a_N(e_2) = \langle y, z \rangle$
- ightharpoonup $lab_N(e_1) = A$, $lab_N(e_2) = B$

The concrete network {y} ► N:



Soft Constraint Satisfaction Problems

- A Soft Constraint Satisfaction Problem (SCSP) is a tuple (I ► N, D, S, val)
- ► A value val_A is a function giving a cost in S to each assignement of canonical variables in A
 - ▶ We will use val_e as a function giving a cost to every assignement to variables e is attached to
- Variables I are those of interest
- ▶ The solution is a function $sol: (I \to \mathbb{D}) \to S$: for each assignement $\rho: I \to \mathbb{D}$

$$sol(
ho) = igoplus_{\{
ho': V_N o \mathbb{D} |
ho' \downarrow_l =
ho\}} ig(igotimes_{\{i | e_i \in E_N\}} val_e(
ho' \downarrow_{a_N(e_1)}) ig)$$

A canonical representative for networks

- ➤ Since we are interested only in interface variables, we can define isomorphisms between networks
- ► Then two networks are isomorphic if there is an isomorphism between them preserving the interface variables
- An abstract network I ▷ N is an isomorphism class of concrete networks
- ► In the following we assume the choice of a canonical representative of abstract networks

Networks and tree decomposition

Concrete and abstract networks

Tree decomposition

- ► A decomposition of a graph can be represented as a tree decomposition
 - Each vertex is a piece of the graph
- ▶ A (rooted) tree decomposition of a hypergraph G is a pair $\mathcal{T} = (\mathcal{T}, X)$ where \mathcal{T} is a rooted tree and $X = \{X_t\}_{t \in V_{\mathcal{T}}}$ is a family of subset of V_G s.t.
 - ▶ For each vertex $v \in V_G$ there exists a vertex of T s.t. $v \in X_t$
 - ► For each hyperedge $e \in E_G$ there is a vertex of \mathcal{T} s.t. $a_G(e) \subseteq X_t$
 - **Each** subtree generated from $v \in V_G$ is a rooted tree

Tree decomposition

- ▶ The decomposition of a network $I \triangleright N$ is a decomposition of N where we choose as a root a vertex of \mathcal{T} containing all the interface variables
- We'll provide a translation from tree decompositions to SCEP (weak) terms: this will enable applying algebraic techniques to tree decompositions
- ightharpoonup We'll refer to *completed versions* of tree decompositions, which explicitly associate components of N to vertices of $\mathcal T$
 - ▶ for each $v \in V_N(e \in E_N)$, $t_v(t_e)$ is the vertex closest to the root of \mathcal{T} s.t. $v \in X_{t_v}$

SCEP signature

- ► SCEP algebras are *permutation algebras*
- ► The SCEP-signature (*s-signature*) is given by the following grammar:

$$p,q:=p\mid\mid q\mid(x)p\mid p\pi\mid A(\widetilde{x})\mid nil$$

- ▶ The free variables of p (fv(p)) are defined recursevly as expected
 - \triangleright the restriction (x) is a bounding operator
 - ▶ the free variables of $p\pi$ are just $\pi(fv(p))$

Strong SCEP specification

- ► A certain set of axioms toghether with the signature give the strong SCEP specification (*s-specification*)
 - ▶ || forms a commutative monoid
 - Restrictions can be swapped and α -converted
 - Permutations distribute over syntactic operators

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 \begin{split} (\mathbf{A}\mathbf{X}_{\parallel}) \\ p \parallel q \equiv_s q \parallel p & (p \parallel q) \parallel r \equiv_s p \parallel (q \parallel r) \\ \mathbf{A}\mathbf{X}_{(x)}) \\ (x)(y)p \equiv_s (y)(x)p & (x)\mathrm{nil} \equiv_s \mathrm{nil} \\ (x)p \equiv_s (y)p[x \mapsto y] & (y \notin fv(p)) \\ (\mathbf{A}\mathbf{X}_{SE}) \\ (x)(p \parallel q) \equiv_s (x)p \parallel q & (x \notin fv(q)) \\ \mathbf{A}\mathbf{X}_{\pi}) \\ (x)(p \parallel q) \equiv_s (x)p \parallel q & (x \notin fv(q)) \\ \mathbf{A}\mathbf{X}_{\pi}) \\ \mathbf{A}(x_1, \dots, x_n)\pi \equiv_s A(\pi(x_1), \dots, \pi(x_n)) \\ \mathbf{nil} \pi \equiv_s \mathrm{nil} & (p \parallel q)\pi \equiv_s p\pi \parallel q\pi \\ ((x)p)\pi \equiv_s (\pi(x))(p\pi) \end{split}
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Weak SCEP specification

- ► The axiom AX_{SE} in the s-specification states that the scope of restricted variables can be narrowed to terms where they occur free
- ► This cause s-terms with different decompositions to be equivalent
- ► To distinguish different decompositions, we define a *weak* SCEP specification (*w-specification*) where *AX_{SE}* is replaced with

$$(AX_{(x)}^w)$$
 $(x)p \equiv_w p$ $(x \notin fv(p))$

Normal and canonical form

- W-terms can be seen as networks having a hierarchical structure
- Normal and canonical forms are those of interest
- A w-term is said to be in normal form whenever it is of the form $(\widetilde{x})(A_1(\widetilde{x_1}) \mid\mid ... \mid\mid A_n(\widetilde{x_n}))$
- A w-term is said to be in canonical form whenever is obtained by the repeated application of (AX_{SE}) until termination
- ► As we'll see, a term's form have an impact on its evaluation complexity

Soundness and completeness of networks

- ▶ The s-specification is sound and complete w.r.t. networks
- We can define a translation function which, given a concrete network I ► N, gives us the corresponding s-term and viceversa
 - $\blacktriangleright term(I \blacktriangleright N) = (V_N \setminus I)(A_1(\widetilde{x_1}) \mid\mid ... \mid\mid A_n(\widetilde{x_n}))$
 - ightharpoonup $net(p) = fv(p)
 ightharpoonup N_p$
- ▶ Completeness follows by the fact that if $net(n_1) \cong net(n_2)$, then $n_1 \equiv_s n_2$

SCSPs as SCEPs

- ► SCSPs can be represented and solved as SCEPs
- ► SCEPs are indeed more general than SCSPs
 - ► There exist optimisation problems which are SCEPs and cannot be represented as SCSPs
- ► Thus we can define an algebra V s.t. $[I \triangleright N]^V$ gives the solution of the SCSP defined over the network
- Theorem Given an SCSP with underlying network $I \triangleright N$ and value functions val_A , we have that $I \triangleright N$ evaluated in $\mathcal{V}(\llbracket I \triangleright N \rrbracket^{\mathcal{V}})$ is its solution

Evaluation Complexity

$$\begin{split} \langle\!\langle p \parallel q \rangle\!\rangle &= \max{\{\langle\!\langle p \rangle\!\rangle, \langle\!\langle q \rangle\!\rangle, |fv(p \parallel q)|\}} &\qquad \langle\!\langle (x)p \rangle\!\rangle = \langle\!\langle p \rangle\!\rangle &\qquad \langle\!\langle p\pi \rangle\!\rangle = \langle\!\langle p \rangle\!\rangle \\ &\qquad \langle\!\langle A(\tilde{x}) \rangle\!\rangle = |set(\tilde{x})| &\qquad \langle\!\langle \text{nil} \rangle\!\rangle = 0 \end{split}$$

- lacktriangle We now introduce a notion of complexity of w-terms $(\langle\langle \rangle\rangle)$
- ▶ The complexity of $p\left(\langle\langle p\rangle\rangle\right)$ is the maximum "size" of elements of an algebra $\mathcal A$ computed while inductively constructing $\llbracket p\rrbracket^{\mathcal A}$

Theorem Given a term p, let n be its normal form. For all canonical forms c of p we have $\langle\langle c \rangle\rangle \leq \langle\langle n \rangle\rangle$

Networks and tree decomposition

Evaluation Complexity

Tree decompositions as w-terms

- ▶ Given a network $I \triangleright N$, let $\mathcal{CT} = (\mathcal{T}, \{t_x\}_{x \in E_N \cup V_N})$ the completed version of one of its tree decompositions
- ightharpoonup We translate \mathcal{CT} into a w-term:
 - ▶ Given $t \in T$, let $V(t) = \{v \in V_N \mid t_v = t\}$ and $E(t) = \{e \in E_N \mid t_e = t\}$
 - Suppose t has children $t_1, ..., t_n$ and E(t) $\{e_1, ..., e_k\}$
 - Let $\widetilde{x} = V(t) \setminus I$
- ▶ The term $\chi(t)$ is inductively defined as:

$$\chi(t) = (\widetilde{x})(A_1(\widetilde{x_1}) \parallel \ldots \parallel A_k(\widetilde{x_k}) \parallel \chi(t_1) \parallel \ldots \parallel \chi(t_n))$$

Tree decompositions as w-terms

- ▶ Given a tree decomposition of \mathcal{T} rooted in r, the corresponding w-term $wterm(\mathcal{T})$ is $\chi(r)$ computed on the completed version of \mathcal{T}
- ▶ $wterm(\mathcal{T})$ correctly represents the network \mathcal{T} decomposes: if \mathcal{T} is a tree decomposition for $I \triangleright N$, then $\llbracket wterm(\mathcal{T}) \rrbracket^{\mathcal{N}} = I \triangleright N$
- ▶ Furthermore, given a tree decomposition \mathcal{T} , we have $\langle \langle \textit{wterm}(\mathcal{T}) \rangle \rangle \leq \textit{width}(\mathcal{T})$

An algorithm for computing canonical decomposition

- Based on bucket elimination
 - Differently from it, the introduced algorithm produces all and only canonical terms
- ▶ Here putting a constraint in the bucket of its last variable means to apply the (AX_{SE}) axiom
- ► The input are an s-term (R)A in normal form and a total order O_R over R
- ▶ The output is a w-term *P* in canonical form

Theorem C is a canonical form of R(A) if and only if there is O_R^C s.t. the algorithm with inputs (R)A and O_R^C outputs C

An algorithm for computing canonical decomposition

Inputs: s-term (R)A in normal form; a total order O_R over R. **Output:** w-term P in canonical form.

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 \begin{array}{ll} 1 & P \leftarrow (R)A \\ 2 & \textbf{while} \ O_R \neq \emptyset \\ 3 & x \leftarrow \text{extract } \max O_R \\ 4 & O_R \leftarrow O_R \setminus \{x\} \\ 5 & \text{find all terms } A' \subseteq A \text{ such that } x \in fv(A') \\ 6 & \textbf{if } A' = \{(R')P'\} \text{ where } P' \text{ has no top-level restriction} \\ 7 & Q \leftarrow \text{ call the algorithm on } (x)P' \text{ with order } \{(x,x)\} \\ 8 & P'' \leftarrow (R')Q \\ 9 & \textbf{else } P'' \leftarrow (x)A' \\ 10 & P \leftarrow (R \setminus \{x\})A \setminus A' \cup \{P''\} \\ 11 & \textbf{return } P \end{array}
```

PSC as an interpretation of the initial algebra

- lacktriangle We define the algebra ${\cal P}$ of polyadic soft constraints
- Constants are $A^{\mathcal{P}}(x_1,...,x_n)\eta=c_{(x_1,...,x_n)}(\eta\circ\widehat{\sigma}), \ \ \textit{nil}^{\mathcal{P}}=0$
- Operations are
 - $((X)^{\mathcal{P}}c)\eta = (\exists_X c)\eta = \bigvee_{\rho} \{c\rho \mid \eta_{|_{\mathbb{V}\setminus X}} = \rho_{|_{\mathbb{V}\setminus X}}\}$
 - $ightharpoonup c\pi^{\mathcal{P}}=(s_{\pi}c)\eta=c(\eta\circ\pi)$
- $ightharpoonup \widehat{\sigma}$ is used to map var(A) to $\langle x_1,...,x_n\rangle$

Weak and strong axioms for ${\mathcal P}$

- ightharpoonup Axioms for lpha-conversion and swapping of restrictions hold in \mathcal{P} , as well as those involving permutations and parallel composition
- $ightharpoonup \mathcal{P}$ is a w-algebra: it holds $(\exists_X c)\eta = c\eta$ if $X \cap supp(c) = \emptyset$ $(AX^w_{(x)})$
- ▶ \mathcal{P} is an s-algebra: it holds $X \cap supp(c2) = \emptyset \implies (\exists_X (c_1 \otimes c_2))\eta = (\exists_X (c_1 \otimes \exists_X c_2))\eta = (\exists_X c_1 \otimes \exists c_2)\eta = (\exists_X c_1)\eta \otimes (\exists_X c_2)\eta = (\exists_X c_1)\eta \otimes c_2\eta$

Example

- Consider a network where meeting activities for a group require the existence of paths between every pair of collaborators
- ► We assume that the network is composed of end-to-end two-way connections with independent probabilities of failure
- We want to find the probability of a certain group to stay connected
- ► We consider a graph with probabilities associated to each edge: the solution is the probability of some interface vertices staying connected

Example

- ▶ We evaluate networks $I \triangleright N$ into an algebra of probability distributions on the partitions of I(Part(I))
- If we are interested in a group J, we compute the probability distribution P for $J \triangleright N$ and then we select $P(\{J\})$
- Let us focus on parallel composition, defined as:

$$\llbracket I_1 \triangleright N_1 \mid \mid I_2 \triangleright N_2 \rrbracket^{\mathcal{D}} \Pi = \sum_{\Pi' \in Part(I \cup x) \mid \Pi' - x = \Pi} \llbracket I_1 \triangleright N_1 \rrbracket^{\mathcal{D}} \Pi_1 \times \llbracket I_2 \triangleright N_2 \rrbracket^{\mathcal{D}} \Pi_2$$

▶ Here $\Pi \in Part(I_1 \cup I_2)$ and each Π_1 , P_2 must belong to $Part(I_1)$ and $Part(I_2)$ respectively. The \cup operator produces the finest partition coarser than the two components and \times is the multiplication on reals

Formal specification

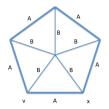
- ► Considered networks are wheels $W_k(v,x)$
- For k = 2 we have the following graph:

$$R_0(x, y, z) = A(x, y) \parallel B(x, z)$$

$$R_{i+1}(x, y, z) = (v)(R_i(x, v, z) \parallel R_i(v, y, z))$$

$$W_k(v, x) = (z)(R_k(x, v, z) \parallel A(v, x) \parallel B(v, z))$$

$$FW_k(v, x) = (z)(R_k(x, v, z) \parallel B(v, z))$$



Non existence of an SCSP formulation

- The given SCEP does not fit the SCSP format
- Suppose to define the problem as an SCSP:
 - a partition in Part(I) can be represented as an assignment of variables I
 - the solution is the probabilty $sol(\Pi)$ associated to a partition Π
- ► The solution in the SCSP case, for any two networks, is $val_{N_1}(\Pi_1) \otimes val_{N_2}(\Pi_2)$
 - The solution considers only the probabilities caused by the same Π on the two subnetworks
- ► This is not compatible with the given definition of parallel composition

References

- ► U. Montanari, M. Sammartino, A. Tcheukam Decomposition Structures for Soft Constraint Evaluation Problems: An Algebraic Approach
- F. Bonchi, L. Bussi, F. Gadducci, F. Santini Polyadic Soft Constraints

Conclusions

- We saw an algebraic framework for defining and solving SCEPs
- We also saw as polyadic soft constraints can be seen as an s-algebra or a w-algebra
- Using these frameworks, we can apply dynamic programming to SCSPs, thus improving their complexity