

InfoCF-Web: An Online Tool for Nonmonotonic Reasoning with Conditionals and Ranking Functions

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Abstract

InfoCF-Web provides implementations of system P and system Z inference, and of inference relations based on c-representation with respect to various inference modes and different classes of minimal models. It has an easy-to-use online interface for computing ranking models of a conditional knowledge \mathcal{R} , and for answering queries and comparing inference results of nonmonotonic inference relations induced by \mathcal{R} .

1 Introduction

Reasoning with conditionals encoding plausible rules of the form “If A, then usually B” has been an area of interest in AI for a long time. While many different semantic approaches have been proposed for dealing with conditionals (e.g. [Lewis, 1973; Kraus *et al.*, 1990; Benferhat *et al.*, 1999]), the actual implementation of the resulting nonmonotonic inference relations seem to have attracted less attention. InfoCF-Web is a system that in addition to providing implementations of system P [Adams, 1975] and system Z [Pearl, 1990] inference, realizes various inference relations based on c-representations [Kern-Isberner, 2001] as models for conditional knowledge bases. It supports several modes of inference and different notions of minimality. The main objective of InfoCF-Web is to easily enable experiments regarding model computations, query answering, and comparison of inference results for small knowledge bases via a web interface. InfoCF-Web is the only currently available experimentation platform of its kind. For working with e.g. multiple or larger knowledge bases and queries, the Java library InfoCF-Lib underlying InfoCF-Web is available.

2 Background: Conditionals and OCFs

Let \mathcal{L} be a propositional language over a finite signature Σ . We write AB for $A \wedge B$ for formulas $A, B \in \mathcal{L}$. We denote the set of all interpretations over \mathcal{L} as Ω . For $\omega \in \Omega$, $\omega \models A$ means that $A \in \mathcal{L}$ holds in ω . We define the set $(\mathcal{L} | \mathcal{L}) = \{(B|A) \mid A, B \in \mathcal{L}\}$ of *conditionals* over \mathcal{L} . The intuition of a conditional $(B|A)$ is that if A holds then usually B holds, too. As semantics for conditionals, we use functions $\kappa : \Omega \rightarrow \mathbb{N}$ such that $\kappa(\omega) = 0$ for at least

one $\omega \in \Omega$, called *ordinal conditional functions (OCF)*, introduced (in a more general form) in [Spohn, 1988]. They express degrees of plausibility of possible worlds where a lower degree denotes “less surprising”. Each κ uniquely extends to a function mapping sentences to $\mathbb{N} \cup \{\infty\}$ given by $\kappa(A) = \min\{\kappa(\omega) \mid \omega \models A\}$ where $\min \emptyset = \infty$. An OCF κ *accepts* a conditional $(B|A)$, written $\kappa \models (B|A)$, if $\kappa(AB) < \kappa(A\bar{B})$. This can also be understood as a non-monotonic inference relation where A κ -*entails* B , written $A \vdash^\kappa B$, if κ accepts $(B|A)$; formally, this is given by

$$A \vdash^\kappa B \text{ iff } A \equiv \perp \text{ or } \kappa(AB) < \kappa(A\bar{B}). \quad (1)$$

A finite set $\mathcal{R} \subseteq (\mathcal{L} | \mathcal{L})$ of conditionals is called a *knowledge base*. An OCF κ accepts \mathcal{R} , written $\kappa \models \mathcal{R}$, if κ accepts all conditionals in \mathcal{R} , and \mathcal{R} is *consistent* if an OCF accepting \mathcal{R} exists [Goldszmidt and Pearl, 1996].

3 Main Features of InfoCF-Web

(1) System P The axiom system P [Adams, 1975; Kraus *et al.*, 1990] provides an important standard for plausible, non-monotonic inferences. B is a *system P inference* from A in the context of \mathcal{R} , written $A \vdash_{\mathcal{R}}^p B$, if B can be derived from the conditionals in \mathcal{R} using the axioms of system P. This holds iff $A \vdash^\kappa B$ for all $\kappa \models \mathcal{R}$ [Kraus *et al.*, 1990].

(2) System Z is based on the ranking function κ^Z , which is the unique Pareto-minimal OCF that accepts \mathcal{R} [Pearl, 1990]. A conditional $(B|A)$ is *tolerated* by a set of conditionals \mathcal{R} if there is a world $\omega \in \Omega$ such that $\omega \models AB$ and $\omega \models \bigwedge_{i=1}^n (\bar{A}_i \vee B_i)$. The definition of κ^Z relies on the unique *inclusion-maximal partition* of \mathcal{R} , denoted by $OP(\mathcal{R}) = (\mathcal{R}_0, \dots, \mathcal{R}_k)$, which is the ordered partition of \mathcal{R} where each \mathcal{R}_i is the (with respect to set inclusion) maximal subset of $\bigcup_{j=i}^k \mathcal{R}_j$ that is tolerated by $\bigcup_{j=i}^k \mathcal{R}_j$ [Goldszmidt and Pearl, 1996]. B can be inferred from A by system Z in the context of \mathcal{R} iff $A \vdash^{\kappa^Z} B$ holds.

(3) C-Representations Other than system Z, the approach of c-representations does not use the most severe falsification of a conditional, but assigns an individual impact to each conditional and generates the world ranks as a sum of impacts of falsified conditionals. For an in-depth introduction to c-representations and their use of the principle of conditional preservation ensured by respecting conditional structures, we refer to [Kern-Isberner, 2001; Kern-Isberner, 2004].

Definition 1 (c-representation [Kern-Isberner, 2001]). A c-representation of a knowledge base \mathcal{R} is a ranking function $\kappa_{\vec{\eta}}$ constructed from $\vec{\eta} = (\eta_1, \dots, \eta_n)$ with integer impacts $\eta_i \in \mathbb{N}_0, i \in \{1, \dots, n\}$ assigned to each conditional $(B_i|A_i)$ such that κ accepts \mathcal{R} and is given by:

$$\kappa_{\vec{\eta}}(\omega) = \sum_{\substack{1 \leq i \leq n \\ \omega \models A_i \bar{B}_i}} \eta_i \quad (2)$$

While for each consistent \mathcal{R} , the system Z ranking function κ^Z is uniquely determined, there may be many different c-representations of \mathcal{R} , denoted by $\mathcal{O}_c(\mathcal{R})$. They can be characterized as the solutions of a constraint satisfaction problem $CR(\mathcal{R})$ [Kern-Isberner, 2004; Beierle et al., 2018].

Definition 2 ($CR(\mathcal{R})$). Let $\mathcal{R} = \{(B_1|A_1), \dots, (B_n|A_n)\}$. The constraint satisfaction problem for c-representations of \mathcal{R} , denoted by $CR(\mathcal{R})$, on the constraint variables $\{\eta_1, \dots, \eta_n\}$ ranging over \mathbb{N}_0 is given by the constraints, for all $i \in \{1, \dots, n\}$:

$$\eta_i > \min_{\omega \models A_i \bar{B}_i} \sum_{\substack{j \neq i \\ \omega \models A_j \bar{B}_j}} \eta_j - \min_{\omega \models A_i \bar{B}_i} \sum_{\substack{j \neq i \\ \omega \models A_j \bar{B}_j}} \eta_j \quad (3)$$

A solution of $CR(\mathcal{R})$ is an n -tuple $(\eta_1, \dots, \eta_n) \in \mathbb{N}_0^n$. For a constraint satisfaction problem CSP , the set of solutions is $Sol(CSP)$. It has been shown that $CR(\mathcal{R})$ is sound and complete, i.e., $\mathcal{O}_c(\mathcal{R}) = \{\kappa_{\vec{\eta}} \mid \vec{\eta} \in Sol(CR(\mathcal{R}))\}$ [Kern-Isberner, 2004], see also [Beierle et al., 2018].

(4) Modes of Inference As pointed out in [Makinson, 1994], historically there have been three different viewpoints when working with multiple plausible models for a knowledge base \mathcal{R} : the choice perspective, focusing on a selected model, the skeptical perspective, considering the intersection of all models, and the credulous perspective, considering the union of all models. System Z represents the choice perspective by identifying a single ranking model for every consistent knowledge base. In [Beierle et al., 2016b], the notion of weakly skeptical inference is introduced. Here, we formalize these different modes of inference over any set of OCFs.

Definition 3 (inference modes over M). Let \mathcal{R} be a knowledge base, M be a set of ranking models of \mathcal{R} , and let A and B be formulas. Then B is a skeptical (credulous, or weakly skeptical, respectively) inference over M in the context of \mathcal{R} from A , denoted by $A \sim_{\mathcal{R}}^{sk, M} B$ ($A \sim_{\mathcal{R}}^{cr, M} B$, or $A \sim_{\mathcal{R}}^{ws, M} B$, respectively) if the following holds:

- $A \sim_{\mathcal{R}}^{sk, M} B$, if $A \sim^{\kappa} B$ for all $\kappa \in M$.
- $A \sim_{\mathcal{R}}^{cr, M} B$, if there is a $\kappa \in M$ such that $A \sim^{\kappa} B$.
- $A \sim_{\mathcal{R}}^{ws, M} B$, if $A \equiv \perp$, or there is a $\kappa \in M$ such that $A \sim^{\kappa} B$ and there is no $\kappa \in M$ such that $A \sim^{\kappa} \bar{B}$.

(5) C-Inference Every single c-representation induces a non-monotonic inference relation as defined in (1). Applying the different inference modes to the set of all c-representations yields the notion of c-inference [Beierle et al., 2016a]. Then B is a skeptical (credulous, or weakly skeptical, respectively) c-inference in the context of \mathcal{R} from A , denoted by

$A \sim_{\mathcal{R}}^{sk, c} B$ ($A \sim_{\mathcal{R}}^{cr, c} B$, or $A \sim_{\mathcal{R}}^{ws, c} B$, respectively), if $A \sim_{\mathcal{R}}^{sk, M} B$ ($A \sim_{\mathcal{R}}^{cr, M} B$, or $A \sim_{\mathcal{R}}^{ws, M} B$, respectively) with $M = \mathcal{O}_c(\mathcal{R})$.

(6) Minimal models While c-representations provide an excellent basis for model-based inference [Kern-Isberner, 2002; Kern-Isberner, 2001], from the point of view of minimal specificity (see e.g. [Benferhat et al., 1992]), those c-representations yielding minimal degrees of implausibility are most interesting [Beierle et al., 2013]. For a knowledge base \mathcal{R} and $\vec{\eta} = (\eta_1, \dots, \eta_n)$ and $\vec{\eta}' = (\eta'_1, \dots, \eta'_n)$ with $\vec{\eta}, \vec{\eta}' \in Sol(CR(\mathcal{R}))$, we define:

$$\vec{\eta} \preceq_+ \vec{\eta}' \quad \text{if} \quad \sum_{1 \leq i \leq n} \eta_i \leq \sum_{1 \leq i \leq n} \eta'_i \quad (4)$$

$$\vec{\eta} \preceq_{cw} \vec{\eta}' \quad \text{if} \quad \eta_i \leq \eta'_i \quad \text{for all } i \in \{1, \dots, n\} \quad (5)$$

$$\vec{\eta} \preceq_O \vec{\eta}' \quad \text{if} \quad \kappa_{\vec{\eta}}(\omega) \leq \kappa_{\vec{\eta}'}(\omega) \quad \text{for all } \omega \in \Omega \quad (6)$$

We write $\vec{\eta} \prec_{\bullet} \vec{\eta}'$ if $\vec{\eta} \preceq_{\bullet} \vec{\eta}'$ and $\vec{\eta}' \not\prec_{\bullet} \vec{\eta}$ for $\bullet \in \{+, cw, O\}$. A vector $\vec{\eta}$ is *sum-minimal* if $\vec{\eta} \preceq_+ \vec{\eta}'$ for all $\vec{\eta}' \in Sol(CR(\mathcal{R}))$; it is *componentwise minimal* (or *cw-minimal*) if there is no vector $\vec{\eta}' \in Sol(CR(\mathcal{R}))$ such that $\vec{\eta}' \prec_{cw} \vec{\eta}$; and it is *ind-minimal* if there is no vector $\vec{\eta}' \in Sol(CR(\mathcal{R}))$ such that $\vec{\eta}' \prec_O \vec{\eta}$.

Thus, while sum-minimal and cw-minimal are defined by just taking the components of the solution vectors $\vec{\eta}$ into account, ind-minimality refers to the OCF induced by $\vec{\eta}$.

(7) Minimal c-inference is obtained by combining the different inference modes with sets of minimal models. Let $\bullet \in \{+, cw, O\}$. Then B is a skeptical (credulous, or weakly skeptical, respectively) \bullet -min-inference in the context of \mathcal{R} from A , denoted by $A \sim_{\mathcal{R}}^{sk, \bullet} B$ ($A \sim_{\mathcal{R}}^{cr, \bullet} B$, or $A \sim_{\mathcal{R}}^{ws, \bullet} B$, resp.), if $A \sim_{\mathcal{R}}^{sk, M, \bullet} B$ ($A \sim_{\mathcal{R}}^{cr, M, \bullet} B$, or $A \sim_{\mathcal{R}}^{ws, M, \bullet} B$, resp.) with $M_{\bullet} = \{\kappa_{\vec{\eta}} \mid \vec{\eta} \in Sol(CR(\mathcal{R})) \text{ and } \vec{\eta} \text{ is } \bullet\text{-minimal}\}$.

Each of the various c-inference relations presented above can be approximated and captured by adding $\eta_i \leq MI$ to the underlying CSP $CR(\mathcal{R})$, thus taking only those impact vectors into account whose impacts are not greater than the *maximal impact* $MI \in \mathbb{N}$ [Beierle and Kutsch, 2019].

4 System Walkthrough and Realisation

InfOCF-Web¹ provides implementations for all inference systems presented in the previous section. Its user interface is shown in Figure 1. In the top left, the user can either manually enter a conditional knowledge base, load a knowledge base from a file, or load a pre-selected demo knowledge base. The syntax is briefly described in the user interface and follows the structure illustrated by the knowledge base \mathcal{R}_{bird} given in Figure 1. To the left of the knowledge base, the user can select ranking models to be calculated. Besides the finite sets of minimal c-representations, the set of all c-representations up to a selected maximal impact can be calculated. The field labelled 'Solutions' then shows the calculated impact vectors, while the top right field will show the actual ranking functions induced by the solutions. If the system Z ranking function is computed, the solutions field will show the ordered partition on which the system Z ranking model is

¹<https://www.fernuni-hagen.de/wbs/research/infocf-web/>

The screenshot displays the InfoOCF-Web interface. At the top, there's a 'Knowledge base' section with a code editor containing logical rules and a 'C-representations / System Z' section with radio buttons for different inference modes. A 'Solutions (2)' section shows a list of solutions. Below this is a table with columns for variables (1., b, f, a, p) and Rank, listing 11 solutions. The bottom section is a 'Query' interface with a query input field containing 'f,b entails a', buttons for 'Answer' and 'Download results', and a table of 'Query results for c-representations'.

1.	b	f	a	p	Rank
1.1.	0	0	0	0	0
1.2.	0	0	0	1	2
1.3.	0	0	1	0	0
1.4.	0	0	1	1	2
1.5.	0	1	0	0	0
1.6.	0	1	0	1	4
1.7.	0	1	1	0	0
1.8.	0	1	1	1	4
1.9.	1	0	0	0	1
1.10.	1	0	0	1	1
1.11.	1	0	1	0	1

query	mode	no	cw-minimal	sum-minimal	ind-minimal	all	maximal impact
(a f,b)	SKEPTICAL	No	No	No	No	No	5
(a f,b)	WEAKLY_SKEPTICAL	Yes	Yes	No	No	Yes	5
(a f,b)	CRECULOUS	Yes	Yes	Yes	No	Yes	5

Figure 1: User interface of InfoOCF-Web.

based. The calculated ranking functions can then be downloaded in a convenient csv-format. For \mathcal{R}_{bird} , there are two cw-minimal c-representations induced by the impact vectors $\vec{\eta}_1 = (1, 1, 0, 2, 2)$ and $\vec{\eta}_2 = (1, 0, 1, 2, 2)$. The first few possible worlds and the ranks under the c-representation $\kappa_{\vec{\eta}_1}$ can be seen in the top right in Figure 1.

The bottom half of the interface is used for querying the inference relations induced by the loaded knowledge base. The user can select multiple sets of c-representations as well as system Z and system P inference. A selected maximal impact is used for inference over sets of c-representations employing the inference modes skeptical, weakly skeptical, and credulous. The query is entered as 'A entails B' for two formulas A and B. Since system P and system Z inference are not affected by the selected inference mode (system P inference is inherently skeptical, while in the case of system Z inference the three inference modes coincide), the query results are displayed next to the query itself. The results for the selected c-inference relations are displayed in a table underneath the query. Figure 1 shows the results for the query 'Does $\bar{f} \wedge b$ entail a?'. The answers determined by InfoOCF-Web for all selected inference systems show that system P and system Z agree on this query, while there are differences among the various c-inference relations; a thorough investigation of properties and interrelationships of all inference methods provided by InfoOCF-Web can be found in [Beierle et al., 2021].

InfoOCF-Web is a web application based on the Java library InfoOCF-Lib [Kutsch, 2019]. System P inference is implemented by computing the ordered partition and thus test-

ing the consistency of the knowledge base augmented by the negated query. System Z inference, as well as minimal c-inference relations, are implemented by calculating the necessary ranking functions and checking the minimal ranks of verification and falsification of the query. While computing OCFs induced by impact vectors for c-representations is done in Java [Kutsch, 2020], the impact vectors are computed as solutions of $CR(\mathcal{R})$ by employing a CSP solver component implemented in a Prolog backend [Beierle et al., 2013]; also answering c-inferences over all c-representations is performed by a Prolog CSP solver by checking the solvability of corresponding CSPs (cf. [Beierle et al., 2018]). Whereas the objective of InfoOCF-Web is to offer a convenient web interface for small scale inference experiments, InfoOCF-Lib offers a broad selection of classes that allow the user to construct conditional knowledge bases, calculate various sets of ranking functions, and solve inference tasks also with respect to sets of knowledge bases and multiple queries; furthermore, sophisticated implementation features such as employing compilation techniques [Beierle et al., 2019] or additional facilities like computing complete inference closures [Kutsch, 2020; Kutsch and Beierle, 2021] are provided.

Acknowledgements

We are grateful to Martin Austen, Alexander Drobel, Fadil Kallat, Matthias Porath, Sandra Schufmann, Leon Schwarzer, Karl Södler, and Matthias Wirths for being involved in creating InfoOCF-Web. This work was partially supported by DFG Grant BE 1700/9-1 awarded to Christoph Beierle.

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