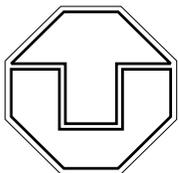


Engineering of Logics for the Content-based Representation of Information

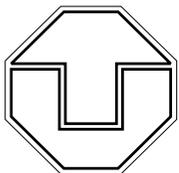
Franz Baader
Theoretical Computer Science
TU Dresden
Germany

- Content-based representation of information
- The role of logics and why they must be engineered
- Description Logics as a successful instance of this approach
- Two applications of DL: Semantic Web and Databases



Content-based representation

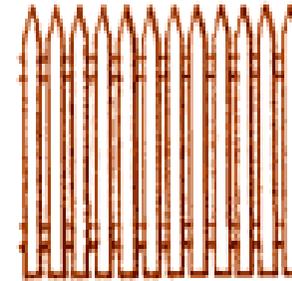
- representation of the "meaning" of the information
- shared understanding of this meaning among all agents (human users, search engines, ...) using the information
- understanding of meaning should result in
 - ability to draw conclusions from the represented information
 - ability to determine semantic equivalence of syntactically different representations



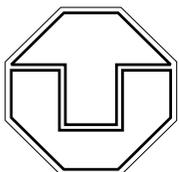
Example

searching for information on the WWW

- looking for garden centers offering **palisades** for my new garden
 - ➔ search engine should know that **paling** is a similar notion
 - ➔ and that **fence** subsumes both



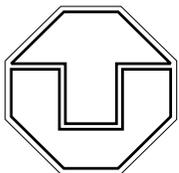
- use of an **ontology**:
 - ➔ defines the important notions (classes, relations, objects) of the domain
 - ➔ states constraints on the way these notions can be interpreted
 - ➔ information about synonyms, subsumption, etc. can automatically be deduced from the definitions and constraints



Semantics

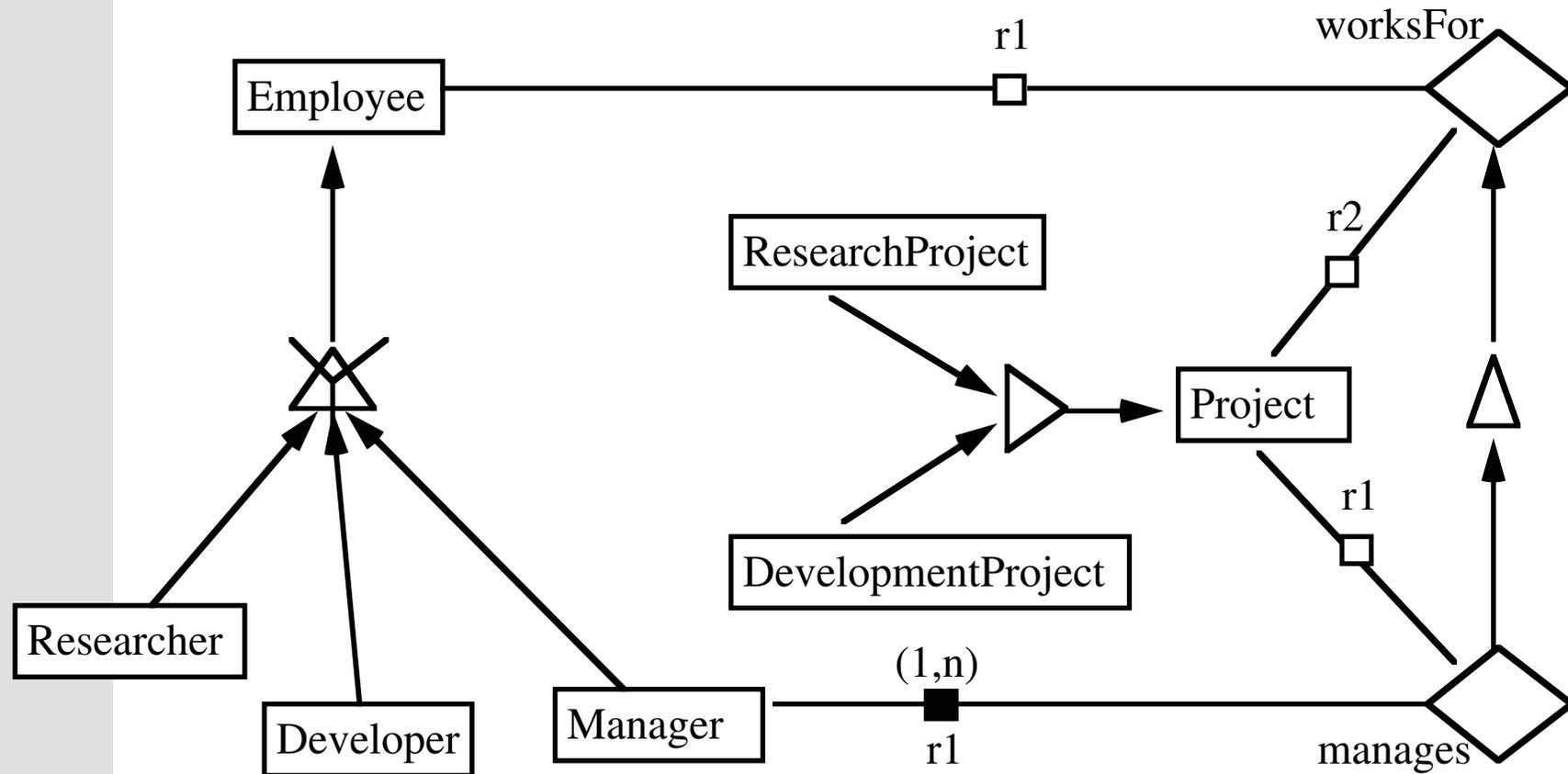
of the representation formalism

- Need for a **formal, well-defined semantics** since otherwise there cannot be a shared understanding and reliable reasoning
 - ➔ not just "intuitive" or purely "procedural" semantics
- **comprehensible** to **human** users
- **usable** by **machines** (e.g. in reasoning)
- **logic** as an appropriate **tool**
 - ➔ yields formal semantics
 - ➔ reasoning about the information as logical inference problem
 - ➔ standard approaches for logical reasoning can be used



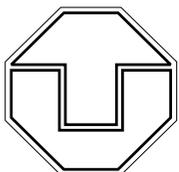
Example

graph-based formalisms such as **semantic networks** (AI),
ER diagrams (DB), **UML diagrams** (software engineering)



Pictures say more than 1000 words,

but they may tell 100 different stories, depending on the viewer.



Why engineering of logics?

- **Expressiveness vs. tractability** issue:

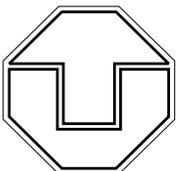
- application-relevant knowledge must be expressible
- reasoning must still be "feasible"

Requires logics that are **tailored** to the application problem

- **Practical considerations, usability** of logics:

not just investigation of formal properties (axiomatization, interpolation, ...), but emphasis on algorithmic properties

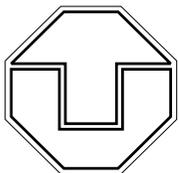
- (worst-case) complexity analysis
- "practical" algorithms
- optimization techniques
- empirical evaluation



Own contributions

to this endeavour

- Designing expressive **knowledge representation** formalisms and practical **reasoning** tools, application in chemical process engineering, databases, and semantic Web
 - ➔ collaboration with E. Franconi, I. Horrocks (U. Manchester), M. Lenzerini (U. Rome), W. Marquardt (RWTH Aachen)
- **Combination** of logics and reasoners: equational theories (word problem and unification), modal and description logics
 - ➔ collaboration with K. Schulz (U. Munich), C. Tinelli (U. Iowa), F. Wolter (U. Leipzig)



Description Logics

class of knowledge representation formalisms

Descended from structured inheritance networks [Brachman 78] via the system KL-ONE [Brachman&Schmolze 85]. Emphasis on well-defined basic inference procedures: subsumption and instance problem.

Phase 1:

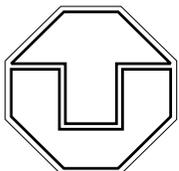
- implementation of incomplete systems (Back, Classic, Loom, ...)
- based on structural subsumption algorithms

Phase 2:

- development of tableau-based algorithms and complexity results
- first implemented tableau-based systems (Kris, Crack)
- first formal investigation of optimization methods

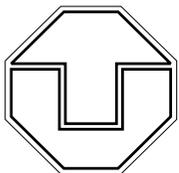
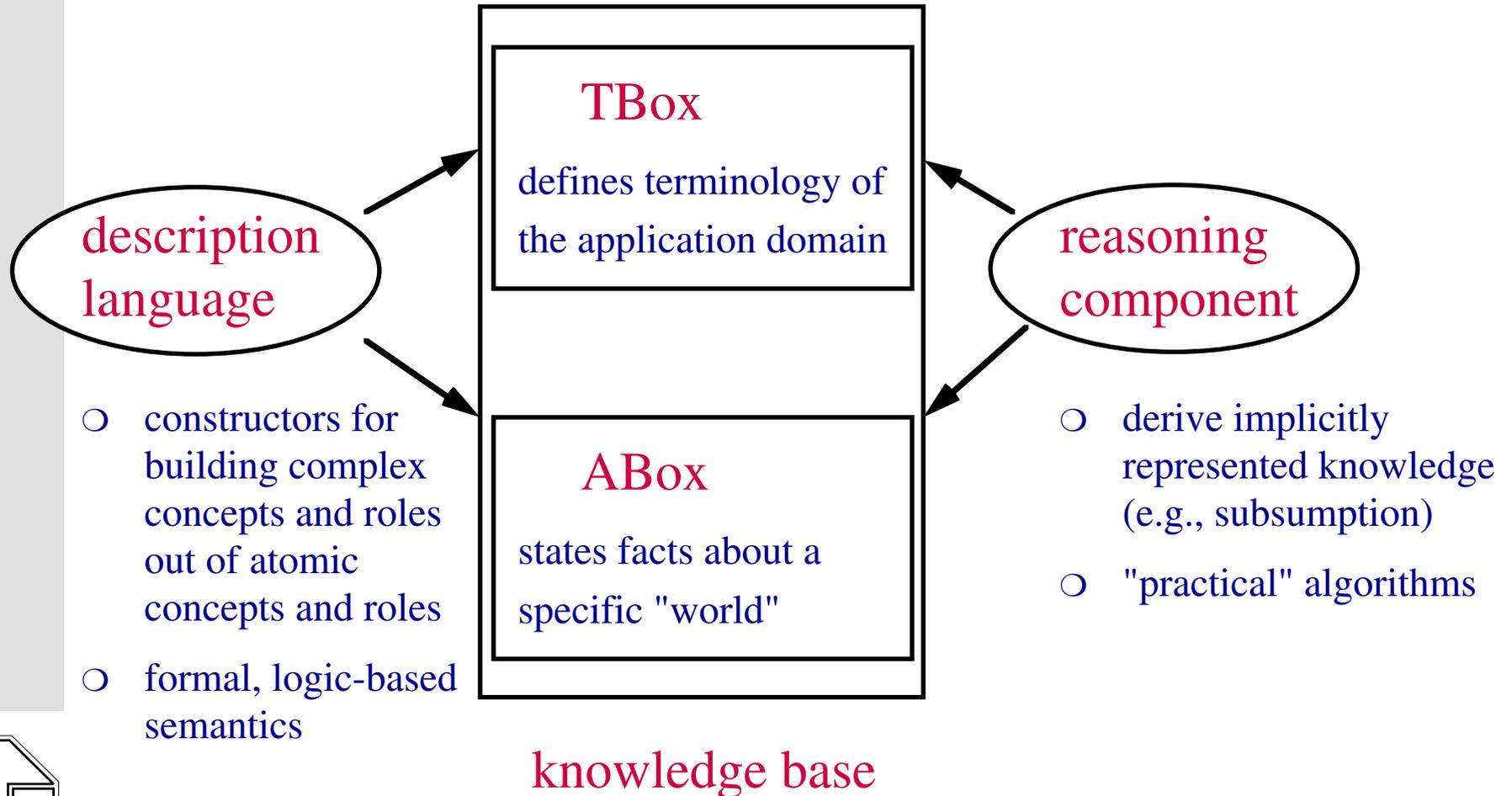
Phase 3:

- tableau-based algorithms for very expressive DLs
- highly optimized tableau-based systems (FaCT, Racer)
- relationship to modal logic and decidable fragments of FOL



Description logic systems

structure



Description language

examples of typical constructors:

$C \sqcap D, \neg C, \forall r. C, \exists r. C, (\geq n r)$

A man

that is married to a doctor, and

has at least 5 children,

all of whom are professors.

$\text{Human} \sqcap \neg \text{Female} \sqcap$

$\exists \text{ married-to} . \text{Doctor} \sqcap$

$(\geq 5 \text{ child}) \sqcap$

$\forall \text{ child} . \text{Professor}$

TBox

definition of concepts

$\text{Happy-man} = \text{Human} \sqcap \dots$

statement of constraints

$\exists \text{ married-to} . \text{Doctor} \sqsubseteq \text{Doctor}$

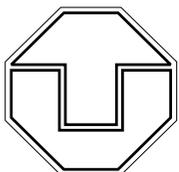
ABox

properties of individuals

$\text{Happy-Man}(\text{Franz})$

$\text{child}(\text{Franz}, \text{Luisa})$

$\text{child}(\text{Franz}, \text{Julian})$



Formal semantics

based on interpretations as in predicate logic

An interpretation I associates

- concepts C with sets C^I and
- roles r with binary relations r^I .

The semantics of the constructors is defined through identities:

- $(C \sqcap D)^I = C^I \cap D^I$
- $(\geq n r)^I = \{d \mid \#\{e \mid (d,e) \in r^I\} \geq n\}$
- $(\forall r. C)^I = \{d \mid \forall e: (d,e) \in r^I \Rightarrow e \in C^I\}$
- ...

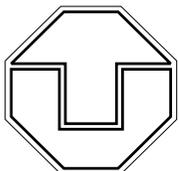
$$I \models A = C \text{ iff } A^I = C^I$$

$$I \models C \sqsubseteq D \text{ iff } C^I \subseteq D^I$$

$$I \models C(a) \text{ iff } a^I \in C^I$$

$$I \models r(a,b) \text{ iff } (a^I, b^I) \in r^I$$

model



Reasoning

makes implicitly represented knowledge explicit,
is provided as service by the DL system, e.g.:

Subsumption: Is C a **subconcept** of D?

$C \sqsubseteq D$ iff $C^I \subseteq D^I$ for all interpretations I.

Satisfiability: Is the concept description C **non-contradictory**?

C is **satisfiable** iff there is an I such that $C^I \neq \emptyset$.

Consistency: Is the ABox \mathcal{A} **non-contradictory**?

\mathcal{A} is **consistent** iff it has a **model**.

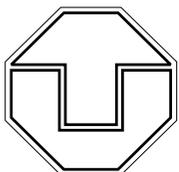
Instantiation: Is e an **instance** of C w.r.t. the given ABox \mathcal{A} ?

$\mathcal{A} \models C(e)$ iff $e^I \in C^I$ for all models I of \mathcal{A} .

*polynomial
reductions*



*in presence
of negation*



Satisfiability algorithm

Idea

generate an interpretation I such that $C_0^I \neq \emptyset$

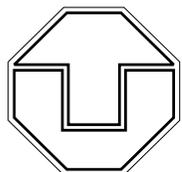
Data structure

for describing (partial) interpretations: **ABoxes**
(w.l.o.g. all concept descriptions in negation normal form)

Approach

ABox assertions are viewed as **constraints**;
propagate constraints.

- Starting with $\mathcal{A}_0 := \{C_0(x_0)\}$, the algorithm applies **transformation rules** until all constraints are satisfied or an obvious contradiction is detected.
- Every **rule corresponds to one constructor**.
- Disjunction requires **non-deterministic** rule: two alternatives.



Exists-restriction rule

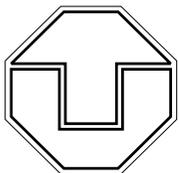
$$\{ \dots \exists r.C(a) \dots \}$$

Condition

there is no c with
 $C(c)$ and $r(a,c)$
present

$$\{ \dots \exists r.C(a), C(b), r(a,b) \dots \}$$

new individual name



Disjunction rule

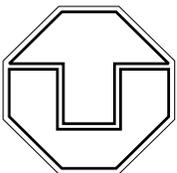
$\{ \dots (C \sqcup D)(a) \dots \}$

Condition

neither
 $C(a)$ nor $D(a)$
is present

$\{ \dots (C \sqcup D)(a), C(a) \dots \}$

$\{ \dots (C \sqcup D)(a), D(a) \dots \}$



search tree

$\{C_0(x_0)\}$

deterministic rule

local soundness: rules preserve satisfiability

[Grey box]

non-deterministic rule

[Grey box]

[Grey box]

termination: all paths finite

⋮

⋮

[Grey box]

complete ABoxes: no rules apply

⋮

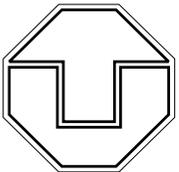
[Grey box]

satisfiable

iff one of the complete ABoxes is open, i.e., does not contain an obvious contradiction (clash)



$A(x), \neg A(x)$



Ontologies

for the Semantic Web

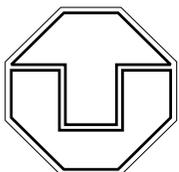
"An ontology is a specification of a conceptualization." (Tom Gruber, Stanford)

- An abstract, simplified view of the world, expressed in an appropriate formal language with well-defined semantics.
- Facilitates **shared understanding**: common ontologies for a set of agents allow them to communicate about a domain of discourse without necessarily operating on a globally shared theory.

DAML+OIL

joint proposal by EU/US initiatives for a W3C ontology standard

- ➔ RDF (schema) based **syntax**
- ➔ **semantics** defined by translation into an expressive DL
- ➔ **reasoning** employs highly optimized DL reasoner (FaCT)

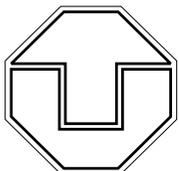


SHIQ

DL used to define the semantics of DAML+OIL
depends on last 10 years of DL research

- **very expressive DL:**
 - ➔ Boolean operators (\sqcap, \sqcup, \neg)
 - ➔ value and existential restrictions ($\forall r.C, \exists r.C$)
 - ➔ qualified number restrictions
 - ➔ general inclusion axioms
 - ➔ transitive roles, inverse roles, and role hierarchies
- **implemented systems:** FaCT [Horrocks 98] and Racer [Haarslev, Moeller 01]
 - ➔ tableau-based subsumption algorithm
building on experience of Kris [B.&Hollunder 91]
 - ➔ highly optimized implementation
building on experience with optimizing Kris
[B., Franconi, Hollunder, Nebel, Profitlich 92]

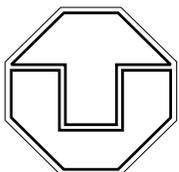
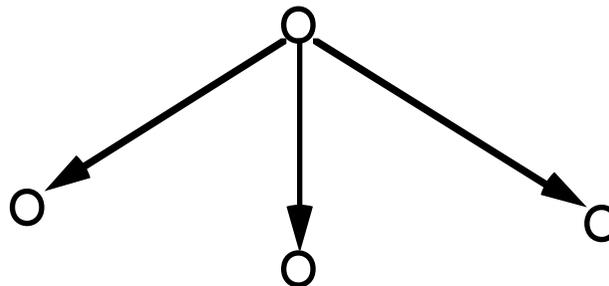
} *ALC*
[Schmidt-Schauß & Smolka 88/91]



Qualified number restrictions

extend the simple number restrictions of early DL systems

- Can not only express "At least 3 children"
(≥ 3 child)
- but also "At most 1 daughter and at most 1 son"
(≤ 1 child.Female) \sqcap (≤ 1 child. \neg Female)
- First algorithm that can handle qualified number restrictions proposed in [Hollunder&B. 91]:
 - Introduces a nondeterministic "choose-rule"
 - necessary to detect inconsistencies:
(≤ 1 child.Female) \sqcap (≤ 1 child. \neg Female) \sqcap (≥ 3 child)

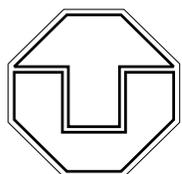


General inclusion axioms

extend the simple concept definitions
of early DL systems

- Can be used to formulate **complex constraints**, e.g.,
 - ➔ domain and range constraints on roles:
 $\exists \text{child} . \text{Human} \sqsubseteq \text{Human}$
 $\text{Human} \sqsubseteq \forall \text{child} . \text{Human}$
- Make **reasoning** considerably **harder** (for \mathcal{ALC} , complexity jumps from PSpace to ExpTime).
- **First algorithm** that can handle general inclusion axioms proposed in [B., Bürckert, Hollunder, Nutt, Siekmann 90]:
 - ➔ termination requires "**blocking**":

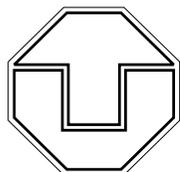
$\text{Human} \sqsubseteq \exists \text{parent} . \text{Human}$



Complex roles

extend the simple atomic roles
of early DL systems

- **Transitive roles** can express paronomies, causality, ..., e.g.,
 $\exists \text{part. (Reactor} \sqcap \exists \text{part. Heater)}$ implies $\exists \text{part. Heater}$
 - Transitive roles in DLs first treated in [Sattler 96]:
ALC with transitive roles still in PSpace.
- **Role hierarchies** can (e.g.) express that **son** is a subrole of **child**
 - Transitive roles and role hierarchies can simulate general inclusion axioms [Horrocks,Sattler 98].
- **Inverse roles**: e.g., **parent** is the inverse of **child**
 - Because of the combination of general inclusion axioms, inverse roles, and number restrictions, *SHIQ* does **not** have the **finite model property**.
 - **First algorithm** for *SHIQ* presented in [Horrocks,Sattler,Tobies 99/00]
 - requires a very **sophisticated blocking** condition.



SHIQ

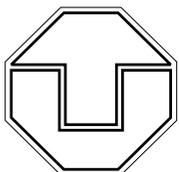
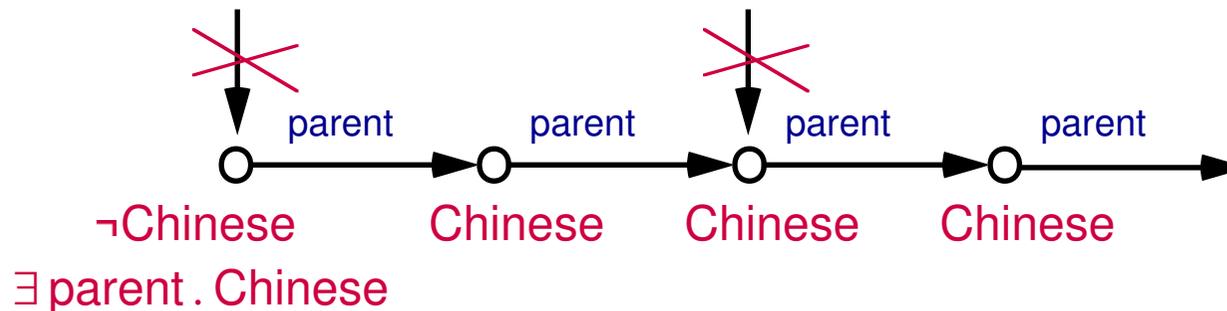
does not have the finite model property

Finite model property: if a subsumption relationship does not hold, then there is a finite counter-model showing this.

Axioms:

$\text{Chinese} \sqsubseteq \exists \text{parent} . \text{Chinese} \sqcap (\leq 1 \text{ child})$
parent is the inverse of child

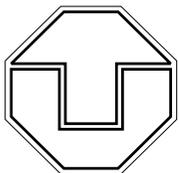
Subsumption question: $\exists \text{parent} . \text{Chinese} \sqsubseteq \text{Chinese} ?$



Conceptual modelling

of data sources

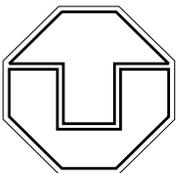
- **Semantic data model** describes the "universe of discourse" about which the database will contain information by
 - introducing the terms to be used in talking about the domain, and
 - capturing their meaning by their inter-relationships and constraints.
- **(Extended) entity-relationship diagrams (EER)** are a semantic modelling formalism that allows to define such models.
- Semantic data models are usually employed in the **design phase**
 - to specify the requirements on the database
 - to generate the logical schema (e.g., in the relational model)
- Semantic data models can also be used
 - when **integrating** different **data sources** (schema integration)
 - for semantic **query optimization**



Description logics

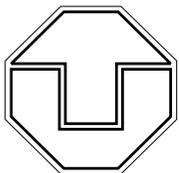
for conceptual modelling

- The DL **DLR** with n-ary relations [Calvanese et al. 99] can express many semantic modelling languages such as EER diagrams.
- The DL *SHIQ* can express the relevant parts of DLR, and thus reasoners for *SHIQ* (like FaCT and Racer) can
 - **check satisfiability** of models expressed in EER
 - **support schema integration** by checking satisfiability of the integrated model
- **ICOM** (Intelligent Conceptual Modelling Tool) [Franconi and Ng 00] realizes this idea.



Conclusion

- Expressive Description Logics can express ontology languages for the Semantic Web and semantic modelling languages for DBs, and provide useful reasoning tools.
- Reasoning in these DLs depends on the last 10 years of DL research
 - justifies our "proactive" research on foundations of DLs
 - which is responsible for the fact that we now have a significant technological lead
- **Future directions:**
 - even more expressive DLs (e.g., practical algorithms for *SHIQ* with individuals)
 - nonstandard inferences in DLs (least common subsumer, matching) that support building and maintaining large ontologies



Overall goal

a warehouse of logics and inference tools

- Offer a **rich palette of logics** with good computational properties.
- Flexible and semantically well-founded **schemes for combining logics** and reasoners.
- Highly **optimized implementations** of reasoning tools.
- Scientifically well-founded **evaluations** in different application domains.

- Achieved by
 - comparing and combining **different reasoning approaches** (automata, tableaux, resolution, BDD, ...)
 - from **different research fields** (automated deduction, knowledge representation, mathematical logic, philosophical logic, verification, ...)

