

E. Description Logics



This section is based on material from

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Description Logics

- OWL DL ist äquivalent zur Beschreibungslogik SHOIM(D_n). Auf letzterer basiert also die Semantik von OWL DL.
- Unter Beschreibungslogiken (Description Logics) versteht man eine Familie von Teilsprachen der Prädikatenlogik 1. Stufe, die entscheidbar sind.
- SHOIN(**D**_n) ist eine relativ komplexe Beschreibungslogik.
- Um einen ersten Einblick in das Prinzip der Beschreibungslogiken zu erhalten, werfen wir zum Abschluss der Vorlesung einen Blick auf etwas abgespeckte Fassungen.

Literatur:

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- F. Baader, W. Nutt: Basic Description Logics. In: Description Logic Handbook, 47-100.
- Ian Horrocks, Peter F. Patel-Schneider and Frank van Harmelen. From SHIQ and RDF to OWL: The making of a web ontology language. http://www.cs.man.ac.uk/%7Ehorrocks/Publications/download/2003/HoPH03a.pdf

Aside: Semantics and Model Theories



- · Ontology/KR languages aim to model (part of) world
- Terms in language correspond to entities in world
- Meaning given by, e.g.:
 - Mapping to another formalism, such as FOL, with own well defined semantics
 - or a bespoke Model Theory (MT)
- MT defines relationship between syntax and interpretations
 - Can be many interpretations (models) of one piece of syntax
 - Models supposed to be analogue of (part of) world
 - · E.g., elements of model correspond to objects in world
 - Formal relationship between syntax and models
 - Structure of models reflect relationships specified in syntax
 - Inference (e.g., subsumption) defined in terms of MT
 - E.g., $\mathcal{T} \models A \sqsubseteq B$ iff in every model of \mathcal{T} , ext(A) \subseteq ext(B)

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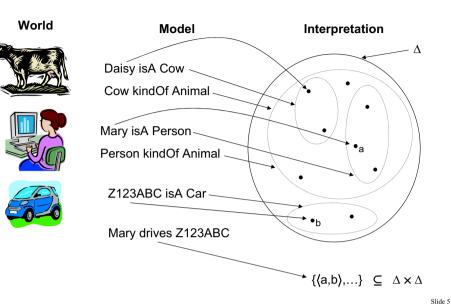
Aside: Set Based Model Theory



- Many logics (including standard First Order Logic) use a model theory based on Zermelo-Frankel set theory
- The domain of discourse (i.e., the part of the world being modelled) is represented as a set (often referred as Δ)
- Objects in the world are interpreted as elements of Δ
 - Classes/concepts (unary predicates) are subsets of Δ
 - Properties/roles (binary predicates) are subsets of $\Delta \times \Delta$ (i.e., Δ^2)
 - Ternary predicates are subsets of Δ^3 etc.
- The sub-class relationship between classes can be interpreted as set inclusion
- Doesn't work for RDF, because in RDF a class (set) can be a member (element) of another class (set)
 - In Z-F set theory, elements of classes are atomic (no structure)



Aside: Set Based Model Theory Example





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Aside: Set Based Model Theory Example

- Formally, the vocabulary is the set of names we use in our model of (part of) the world
 - {Daisy, Cow, Animal, Mary, Person, Z123ABC, Car, drives, ...}
- An interpretation *I* is a tuple (Δ, .^{*I*})
 - Δ is the domain (a set)
 - $\cdot^{\mathcal{I}}$ is a mapping that maps
 - Names of objects to elements of Δ
 - Names of unary predicates (classes/concepts) to subsets of $\boldsymbol{\Delta}$
 - Names of binary predicates (properties/roles) to subsets of $\Delta \times \Delta$
 - · And so on for higher arity predicates (if any)

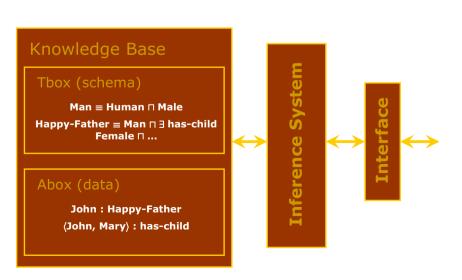




- A family of logic based Knowledge Representation formalisms
 - Descendants of semantic networks and KL-ONE
 - Describe domain in terms of concepts (classes), roles (relationships) and individuals
- Distinguished by:
 - Formal semantics (typically model theoretic)
 - · Decidable fragments of FOL
 - Closely related to Propositional Modal & Dynamic Logics
 - Provision of inference services
 - Sound and complete decision procedures for key problems
 - Implemented systems (highly optimised)

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DL Architecture





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Short History of Description Logics



Phase 1:

- Incomplete systems (Back, Classic, Loom, . . .)
- Based on structural algorithms

Phase 2:

- Development of tableau algorithms and complexity results
- Tableau-based systems for Pspace logics (e.g., Kris, Crack)
- Investigation of optimisation techniques

Phase 3:

- Tableau algorithms for very expressive DLs
- Highly optimised tableau systems for ExpTime logics (e.g., FaCT, DLP, Racer)
- Relationship to modal logic and decidable fragments of FOL

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Latest Developments



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Phase 4:

- Mature implementations
- Mainstream applications and Tools
 - Databases
 - Consistency of conceptual schemata (EER, UML etc.)
 - Schema integration
 - Query subsumption (w.r.t. a conceptual schema)
 - · Ontologies and Semantic Web (and Grid)
 - Ontology engineering (design, maintenance, integration)
 - Reasoning with ontology-based markup (meta-data)
 - Service description and discovery
- Commercial implementations
 - · Cerebra system from Network Inference Ltd

From RDF to OWL



- · Two languages developed to satisfy the requirements
 - OIL: developed by group of (largely) European researchers (several from EU OntoKnowledge project)
 - DAML-ONT: developed by group of (largely) US researchers (in DARPA DAML programme)
- Efforts merged to produce DAML+OIL
 - Development was carried out by "Joint EU/US Committee on Agent Markup Languages"
 - Extends ("DL subset" of) RDF
- DAML+OIL submitted to W3C as basis for standardisation
 - Web-Ontology (WebOnt) Working Group formed
 - WebOnt group developed OWL language based on DAML+OIL
 - OWL language now a W3C Recommendation (i.e., a standard like HTML and XML)

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Description Logic Family



- DLs are a family of logic based KR formalisms
- Particular languages mainly characterised by:
 - Set of constructors for building complex concepts and roles from simpler ones
 - Set of axioms for asserting facts about concepts, roles and individuals
- \mathcal{ALC} is the smallest DL that is propositionally closed
 - Constructors include booleans (and, or, not), and
 - Restrictions on role successors
 - E.g., concept describing "happy fathers" could be written:

Man ∧ ∃hasChild.Female ∧ ∃hasChild.Male ∧ ∀hasChild.(Rich ∨ Happy)

DL Concept and Role Constructors



- Range of other constructors found in DLs, including:
 - Number restrictions (cardinality constraints) on roles, e.g., ≥3 hasChild, ≤1 hasMother
 - Qualified number restrictions, e.g., \geq 2 hasChild.Female, \leq 1 hasParent.Male
 - Nominals (singleton concepts), e.g., {Italy}
 - Concrete domains (datatypes), e.g., hasAge.(< 21)
 - Inverse roles, e.g., hasChild (hasParent)
 - Transitive roles, e.g., hasChild* (descendant)
 - Role composition, e.g., hasParent ∘ hasBrother (uncle)

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DL Knowledge Base



- DL Knowledge Base (KB) normally separated into 2 parts:
 - TBox is a set of axioms describing structure of domain (i.e., a conceptual schema), e.g.:
 - HappyFather \equiv Man \land \exists hasChild.Female \land ...
 - Elephant ≡ Animal ∧ Large ∧ Grey
 - transitive(ancestor)
 - ABox is a set of axioms describing a concrete situation (data), e.g.:
 - · John:HappyFather
 - <John,Mary>:hasChild
- · Separation has no logical significance
 - But may be conceptually and implementationally convenient

OWL as DL: Class Constructors



Constructor	DL Syntax	Example	FOL Syntax
intersectionOf	$C_1 \sqcap \ldots \sqcap C_n$	Human □ Male	$C_1(x) \wedge \ldots \wedge C_n(x)$
unionOf	$C_1 \sqcup \ldots \sqcup C_n$	Doctor ⊔ Lawyer	$C_1(x) \vee \ldots \vee C_n(x)$
complementOf	$\neg C$	¬Male	$\neg C(x)$
oneOf	$ \{x_1\} \sqcup \ldots \sqcup \{x_n\} $	{john} ⊔ {mary}	$ x = x_1 \lor \ldots \lor x = x_n$
allValuesFrom	$\forall P.C$	∀hasChild.Doctor	$\forall y.P(x,y) \rightarrow C(y)$
someValuesFrom	$\exists P.C$	∃hasChild.Lawyer	$\exists y. P(x,y) \land C(y)$
maxCardinality	$\leq nP$	≤1hasChild	$\exists^{\leqslant n} y. P(x,y)$
minCardinality	$\geqslant nP$	≥2hasChild	$\exists^{\geqslant n}y.P(x,y)$

- XMLS datatypes as well as classes in ∀P.C and ∃P.C
 - E.g., ∃hasAge.nonNegativeInteger
- Arbitrarily complex nesting of constructors
 - E.g., Person □ ∀hasChild.Doctor □ ∃hasChild.Doctor

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RDFS Syntax



E.g., Person □ ∀hasChild.Doctor ⊔ ∃hasChild.Doctor:

```
<owl:Class>
  <owl:intersectionOf rdf:parseType=" collection">
    <owl:Class rdf:about="#Person"/>
    <owl:Restriction>
      <owl:onProperty rdf:resource="#hasChild"/>
      <owl:toClass>
        <owl:unionOf rdf:parseType=" collection">
          <owl:Class rdf:about="#Doctor"/>
          <owl:Restriction>
            <owl:onProperty rdf:resource="#hasChild"/>
            <owl:hasClass rdf:resource="#Doctor"/>
          </owl:Restriction>
        </owl:unionOf>
      </owl:toClass>
    </owl:Restriction>
  </owl:intersectionOf>
</owl:Class>
```

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OWL as DL: Axioms



Axiom	DL Syntax	Example
subClassOf	$C_1 \sqsubseteq C_2$	Human ⊑ Animal □ Biped
equivalentClass	$C_1 \equiv C_2$	Man ≡ Human □ Male
disjointWith	$C_1 \sqsubseteq \neg C_2$	Male ⊑ ¬Female
sameIndividualAs	$\{x_1\} \equiv \{x_2\}$	${President_Bush} \equiv {G_W_Bush}$
differentFrom	$\{x_1\} \sqsubseteq \neg \{x_2\}$	$\{\text{john}\} \sqsubseteq \neg \{\text{peter}\}$
subPropertyOf	$P_1 \sqsubseteq P_2$	hasDaughter \sqsubseteq hasChild
equivalentProperty	$P_1 \equiv P_2$	cost ≡ price
inverseOf	$P_1 \equiv P_2^-$	hasChild ≡ hasParent ⁻
transitiveProperty	$P^+ \sqsubseteq \tilde{P}$	ancestor ⁺ ⊑ ancestor
functionalProperty	$\top \sqsubseteq \leqslant 1P$	T ⊑ ≤1hasMother
inverseFunctionalProperty	⊤ ⊑ ≤ 1 <i>P</i> −	\top \sqsubseteq ≤1hasSSN $^-$

- Axioms (mostly) reducible to inclusion (□)
 - $C \equiv D$ iff both $C \sqsubseteq D$ and $D \sqsubseteq C$
- Obvious FOL equivalences
 - E.g., $C \equiv D$ iff $\forall x$. $C(x) \Leftrightarrow D(x)$,
 - $C \sqsubseteq D$ iff $\forall x$. $C(x) \Rightarrow D(x)$

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XML Schema Datatypes in OWL



- OWL supports XML Schema primitive datatypes
 - E.g., integer, real, string, ...
- Strict separation between "object" classes and datatypes
 - Disjoint interpretation domain $\Delta_{\rm D}$ for datatypes
 - For a datavalue d holds $d^{\mathcal{I}} \subseteq \Delta_D$
 - and $\Delta_{\mathrm{D}} \cap \Delta^{\mathcal{I}} = \emptyset$
 - Disjoint "object" and datatype properties
 - For a datatype propterty P holds $P^{\mathcal{I}} \subseteq \Delta^{\mathcal{I}} \times \Delta_{D}$
 - For object property ${\rm S}$ and datatype property ${\rm P}$ hold ${\rm S}^{\mathcal{I}}\cap {\rm P}^{\mathcal{I}}$ = \emptyset
- Equivalent to the " (D_n) " in $\mathcal{SHOIN}(D_n)$

Why Separate Classes and Datatypes?



- Philosophical reasons:
 - Datatypes structured by built-in predicates
 - Not appropriate to form new datatypes using ontology language
- Practical reasons:
 - Ontology language remains simple and compact
 - Semantic integrity of ontology language not compromised
 - Implementability not compromised can use hybrid reasoner

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OWL DL Semantics



- Mapping OWL to equivalent DL $(SHOIN(D_n))$:
 - Facilitates provision of reasoning services (using DL systems)
 - Provides well defined semantics
- DL semantics defined by interpretations: $\mathcal{I} = (\Delta^{\mathcal{I}}, \cdot^{\mathcal{I}})$, where
 - $-\Delta^{\mathcal{I}}$ is the domain (a non-empty set)
 - $\cdot^{\mathcal{I}}$ is an interpretation function that maps:
 - Concept (class) name A to subset $A^{\mathcal{I}}$ of $\Delta^{\mathcal{I}}$
 - Role (property) name R to binary relation $R^{\mathcal{I}}$ over $\Delta^{\mathcal{I}}$
 - Individual name i to element $i^{\mathcal{I}}$ of $\Delta^{\mathcal{I}}$

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DL Semantics



 Interpretation function .^I extends to concept expressions in the obvious way, i.e.:

$$(C \sqcap D)^{\mathcal{I}} = C^{\mathcal{I}} \cap D^{\mathcal{I}}$$

$$(C \sqcup D)^{\mathcal{I}} = C^{\mathcal{I}} \cup D^{\mathcal{I}}$$

$$(\neg C)^{\mathcal{I}} = \Delta^{\mathcal{I}} \setminus C^{\mathcal{I}}$$

$$\{x\}^{\mathcal{I}} = \{x^{\mathcal{I}}\}$$

$$(\exists R.C)^{\mathcal{I}} = \{x \mid \exists y. \langle x, y \rangle \in R^{\mathcal{I}} \land y \in C^{\mathcal{I}}\}$$

$$(\forall R.C)^{\mathcal{I}} = \{x \mid \forall y. (x, y) \in R^{\mathcal{I}} \Rightarrow y \in C^{\mathcal{I}}\}$$

$$(\leqslant nR)^{\mathcal{I}} = \{x \mid \#\{y \mid \langle x, y \rangle \in R^{\mathcal{I}}\} \leqslant n\}$$

$$(\geqslant nR)^{\mathcal{I}} = \{x \mid \#\{y \mid \langle x, y \rangle \in R^{\mathcal{I}}\} \geqslant n\}$$

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DL Knowledge Bases (Ontologies)

- An OWL ontology maps to a DL Knowledge Base $\mathcal{K} = \langle \mathcal{T}, \mathcal{A} \rangle$
 - $-\mathcal{T}$ (Tbox) is a set of axioms of the form:
 - $C \sqsubseteq D$ (concept inclusion)
 - $C \equiv D$ (concept equivalence)
 - $R \sqsubseteq S$ (role inclusion)
 - $R \equiv S$ (role equivalence)
 - $R^+ \sqsubseteq R$ (role transitivity)
 - \mathcal{A} (Abox) is a set of axioms of the form
 - $x \in D$ (concept instantiation)
 - $\langle x,y \rangle \in R$ (role instantiation)
- Two sorts of Tbox axioms often distinguished
 - "Definitions"
 - $C \sqsubseteq D$ or $C \equiv D$ where C is a concept name
 - General Concept Inclusion axioms (GCIs)
 - $C \sqsubseteq D$ where C in an arbitrary concept

Knowledge Base Semantics



- An interpretation \mathcal{I} satisfies (models) an axiom $A (\mathcal{I} \models A)$:
 - $-\mathcal{I} \models \mathcal{C} \sqsubseteq \mathcal{D} \text{ iff } \mathcal{C}^{\mathcal{I}} \subseteq \mathcal{D}^{\mathcal{I}}$
 - $-\mathcal{I} \models C \equiv D \text{ iff } C^{\mathcal{I}} = D^{\mathcal{I}}$
 - $-\mathcal{I} \models R \sqsubseteq S \text{ iff } R^{\mathcal{I}} \subseteq S^{\mathcal{I}}$
 - $-\mathcal{I} \models R \equiv S \text{ iff } R^{\mathcal{I}} = S^{\mathcal{I}}$
 - $-\mathcal{I} \models R^+ \sqsubseteq R \text{ iff } (R^{\mathcal{I}})^+ \subseteq R^{\mathcal{I}}$
 - $\mathcal{I} \models x \in D$ iff $x^{\mathcal{I}} \in D^{\mathcal{I}}$
 - $\mathcal{I} \vDash \langle x, y \rangle \in R \text{ iff } (x^{\mathcal{I}}, y^{\mathcal{I}}) \in R^{\mathcal{I}}$
- \mathcal{I} satisfies a Tbox \mathcal{T} ($\mathcal{I} \models \mathcal{T}$) iff \mathcal{I} satisfies every axiom A in \mathcal{T}
- \mathcal{I} satisfies an Abox \mathcal{A} ($\mathcal{I} \models \mathcal{A}$) iff \mathcal{I} satisfies every axiom A in \mathcal{A}
- \mathcal{I} satisfies an KB \mathcal{K} ($\mathcal{I} \models \mathcal{K}$) iff \mathcal{I} satisfies both \mathcal{T} and \mathcal{A}

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Inference Tasks



- Knowledge is correct (captures intuitions)
 - C subsumes D w.r.t. \mathcal{K} iff for every model \mathcal{I} of \mathcal{K} , $C^{\mathcal{I}} \subseteq D^{\mathcal{I}}$
- Knowledge is minimally redundant (no unintended synonyms)
 - C is equivalent to D w.r.t. \mathcal{K} iff for every model \mathcal{I} of \mathcal{K} , $C^{\mathcal{I}} = D^{\mathcal{I}}$
- Knowledge is meaningful (classes can have instances)
 - C is satisfiable w.r.t. \mathcal{K} iff there exists some model \mathcal{I} of \mathcal{K} s.t. $C^{\mathcal{I}} \neq \emptyset$
- Querying knowledge
 - x is an instance of C w.r.t. \mathcal{K} iff for every model \mathcal{I} of \mathcal{K} , $x^{\mathcal{I}} \in C^{\mathcal{I}}$
 - $\quad \langle x,y \rangle \text{ is an instance of } R \text{ w.r.t. } \mathcal{K} \text{ iff for, } \underline{\textit{every}} \text{ model } \mathcal{I} \text{ of } \mathcal{K} \text{, } (x^{\mathcal{I}}\!,\!y^{\mathcal{I}}\!) \in R^{\mathcal{I}}$
- Knowledge base consistency
 - A KB K is consistent iff there exists *some* model I of K

DL Reasoning



- · Tableau algorithms used to test satisfiability (consistency)
- Try to build a tree-like model I of the input concept C
- Decompose C syntactically
 - Apply tableau expansion rules
 - Infer constraints on elements of model
- Tableau rules correspond to constructors in logic (□, □ etc)
 - Some rules are nondeterministic (e.g., □, ≤)
 - In practice, this means search
- Stop when no more rules applicable or clash occurs
 - Clash is an obvious contradiction, e.g., A(x), $\neg A(x)$
- Cycle check (blocking) may be needed for termination
- C satisfiable iff rules can be applied such that a fully expanded clash free tree is constructed

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Highly Optimised Implementation

- Naive implementation leads to effective non-termination
- Modern systems include MANY optimisations
- Optimised classification (compute partial ordering)
 - Use enhanced traversal (exploit information from previous tests)
 - Use structural information to select classification order
- Optimised subsumption testing (search for models)
 - Normalisation and simplification of concepts
 - Absorption (rewriting) of general axioms
 - Davis-Putnam style semantic branching search
 - Dependency directed backtracking
 - Caching of satisfiability results and (partial) models
 - Heuristic ordering of propositional and modal expansion

- ...

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