INF5390 - Kunstig intelligens

Knowledge Representation

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Outline

- General ontology
- Categories and objects
- Events and processes
- Category reasoning
- Default logics
- Internet shopping world
- Summary

Extracts from AIMA
Chapter 12: Knowledge Representation

Ontologies

- An ontology is a "vocabulary" and a "theory" of a certain "part of reality"
- Special-purpose ontologies apply to restricted domains (e.g. electronic circuits)
- General-purpose ontologies have wider applicability across domains, i.e.
 - ✓ Must include concepts that cover many subdomains
 - √ Cannot use special "short-cuts" (such as ignoring time)
 - Must allow unification of different types of knowledge
- GP ontologies are useful in widening applicability of reasoning systems, e.g. by including time

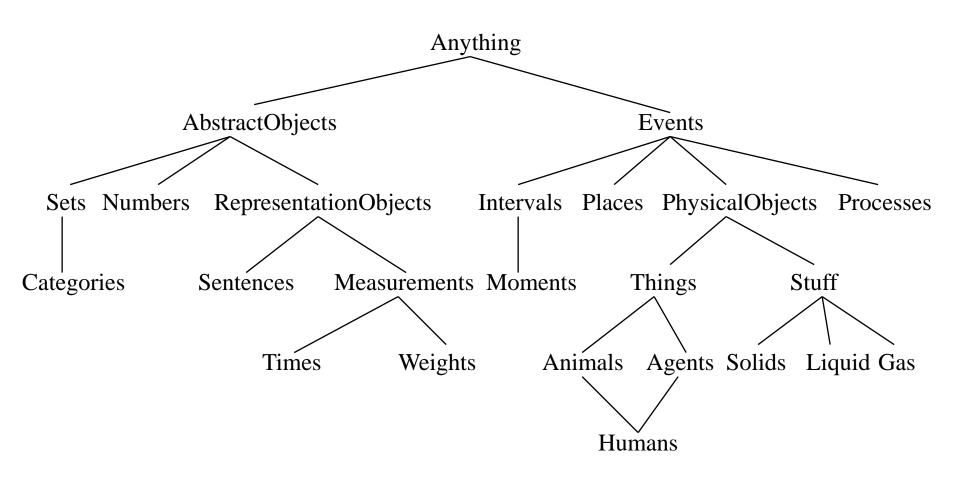
Ontological engineering

- Representing a general-purpose ontology is a difficult task called ontology engineering
- Existing GP ontologies have been created in different ways:
 - By team of trained ontologists
 - By importing concepts from database(s)
 - By extracting information from text documents
 - → By inviting anybody to enter commonsense knowledge
- Ontological engineering has only been partially successful, and few large AI systems are based on GP ontologies (use special purpose ontologies)

Elements of a general ontology

- Categories of objects
- Measures of quantities
- Composite objects
- Time, space, and change
- Events and processes
- Physical objects
- Substances
- Mental objects and beliefs

Top-level ontology of the world



Categories and objects

 Categories are used to classify objects according to common properties or definitions

 $\forall xx \in Tomates \Rightarrow Red(x) \land Round(x)$

- Categories can be represented by
 - \checkmark Predicates: *Tomato*(x)
 - √ Sets: The constant *Tomatoes* represents set of tomatoes (reification)
- Roles of category representations
 - ✓ Instance relations (is-a): $x_1 \in Tomatoes$
 - ✓ Taxonomical hierarchies (*Subset*): $Tomatoes \subset Fruit$
 - √ Inheritance of properties
 - √ (Exhaustive) decompositions

Objects and substance

- Need to distinguish between substance and discrete objects
- Substance ("stuff")
 - √ Mass nouns not countable
 - √ Intrinsic properties
 - ✓ Part of a substance is (still) the same substance
- Discrete objects ("things")
 - √ Count nouns countable
 - Extrinsic properties
 - Parts are (generally) not of same category

Composite objects

- A composite object is an object that has other objects as parts
- The PartOf relation defines the object containment, and is transitive and reflexive

```
PartOf(x, y) \land PartOf(y, z) \Rightarrow PartOf(x, z)

PartOf(x, x)
```

- Objects can be grouped in *PartOf* hierarchies, similar to *Subset* hierarchies
- The structure of the composite object describes how the parts are related

Measurements

- Need to be able to represent properties like height, mass, cost, etc. Values for such properties are measures
- Unit functions represent and convert measures

```
Length(L_1) = Inches(1.5) = Centimeters(3.81)
\forall l \ Centimeters(2.54 \times l) = Inches(l)
```

Measures can be used to describe objects

```
Mass(Tomato_1) = Kilograms(0.16)
\forall dd \in Days \Rightarrow Duration(d) = Hours(24)
```

Non-numerical measures can also be represented, but normally there is an order (e.g. >). Used in qualitative physics

Event calculus

- Event calculus: How to deal with change based on representing points of time
- Reifies fluents (No: "forløp") and events
 - √ A fluent: At(Shankar, Berkeley)
 - √ The fluent is true at time t: T(At(Shankar, Berkeley),t).
- Events are instances of event categories

```
E_1 \in Flyings \land Flyer(E_1, Shankar) \land Origin(E_1, SF) \land Destination(E_1, LA)
```

- Event E₁ took place over interval i
 - √ Happens(E₁, i)
- Time intervals represented by (start, end) pairs

$$\sqrt{i} = (t_1, t_2)$$

Event calculus predicates

- T(f, t) Fluent f is true at time t
- Happens(e, i) Event e happens over interval i
- Initiates(e, f, t) Event e causes fluent f to start at t
- Terminates(e, f, t) Event e causes f to cease at t
- Clipped(f, i)
 Fluent f ceases to be true in int. i
- Restored(f, i)
 Fluent f becomes true in interval i

Using event calculus

- Can extend predicate to cover intervals $T(f,(t_1,t_2)) \Leftrightarrow [\forall t \ (t_1 \leq t < t) \Rightarrow T(f,t)]$
- Fluents and actions are defined with domainspecific axioms, e.g. in the wumpus world
 - $Initiates(e, HaveArrow(a), t) \Leftrightarrow e = Start$ $Terminates(e, HaveArrow(a), t) \Leftrightarrow e \in Shootings(a)$
- Can extend event calculus to represent simultaneous events, continuous events, etc.

Time intervals

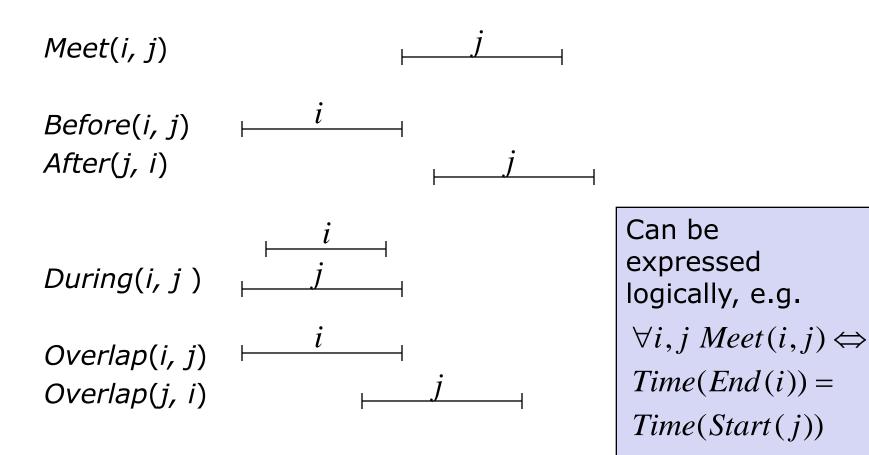
 Time intervals are partitioned into moments (zero duration) and extended intervals

 $Partition(\{Moments, ExtendedIntervals\}, Intervals)$

 $\forall i \ i \in Intervals \Rightarrow (i \in Moments \Leftrightarrow Duration(i) = 0)$

- Functions Start and End delimit intervals
 - $\forall i \ Interval(i) \Rightarrow Duration(i) = (Time(End(i)) Time(Start(i)))$
- May use e.g. January 1, 1900 as arbitrary time 0
 Time(Start(AD1900))=Seconds(0)

Relations between time intervals



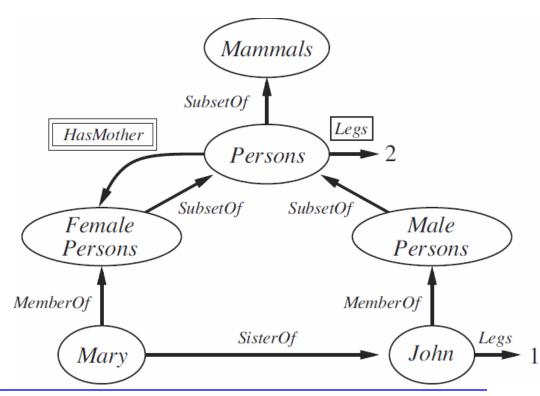
Mental events and mental objects

- Need to represent beliefs in self and other agents, e.g. for controlling reasoning, or for planning actions that involve others
- How are beliefs represented?
 - √ Beliefs are reified as mental objects
 - Mental objects are represented as strings in a language
 - √ Inference rules for this language can be defined.
- Rules for reasoning about logical agents' use their beliefs

```
\forall a, p, q \ LogicalAgent(a) \land Believes(a, p) \land
Believes(a, "p \Rightarrow q") \Rightarrow Believes(a, q)
\forall a, p \ LogicalAgent(a) \land Believes(a, p)
\Rightarrow Believes(a, "Believes(Name(a), p)")
```

Semantic networks

- Graph representation of categories, objects, relations, etc. (i.e. essentially FOL)
- Natural representation of inheritance and default values
 - ✓ E.g. a Person has normally 2 legs, but the default is overridden for John with 1 leg



Description logic (DL)

- FOL enables ascribing properties to objects, while DL allows formal specification of and reasoning about definitions and categories
- DL inference tasks:
 - √ Subsumption Check if a category is a subset of another
 - √ Classification Check if object belongs to a category
 - √ Consistency Check if category definition is satisfiable
- DL evolved from semantic networks as a more formalized approach, still based on taxonomies
- DL in different versions is the logical foundation for the Semantic Web

CLASSIC - DL language

- CLASSIC is an early example of DL, in which definitions can be stated and reasoned about
- Simple category definitions
 - √ Single = And(Unmarried, Adult)
 - √ Bachelor = And(Unmarried, Adult, Male)
- CLASSIC can answer questions like
 - Is category Bachelor subsumed by category Single?
 - √ Is the individual Adam of category Bachelor?
- CLASSIC definitions can be translated to FOL, but inference in DL is more efficient

Defaults and non-monotonic logic

- Classical logic is monotonic: true statements remain true after new facts are added to KB
 ✓ If KB | α, then KB ∧ β | α
- In the closed-world assumption (facts not mentioned assumed false), monotonicity is violated
 - ✓ If α is not mentioned in KB, then $KB \models \neg \alpha$, but $KB \land \alpha \models \alpha$
- Non-monotonic reasoning is widespread in common-sense reasoning
 - We assume default in absence of other input, and are able to retract assumption if new evidence occurs
- Non-monotonic logics support such reasoning

Circumscription (no: "begrensning")

- Circumscription is a more powerful version of the closed-world assumption
 - √ The idea is to specify particular predicates "as false as possible", i.e. false for every object except for those for which they are known to be true
- E.g. for the default that birds can fly
 - **⋄** Bird(x) \land ¬Abnormal(x) \Rightarrow Flies(x)
- If Abnormal is circumscribed, a circumscriptive reasoner can
 - √ Assume ¬Abnormal(x) unless the opposite is known
 - √ Infer Flies(Tweety) from Bird(Tweety)
 - Retract the conclusion if Abnormal(Tweety) is asserted

Truth maintenance systems

- Many inferences in the KB may have default status, and may need to be retracted in a process called **belief revision**
 - √ E.g. KB contains statement P (a default)
 - \checkmark New evidence that P is not true: TELL(KB, \neg P)
 - √ To avoid contradiction: RETRACT(KB, P)
 - ✓ Other statements may have been added by P, e.g. Q if the KB contains $P \Rightarrow Q$, and Q may also have to go
 - ✓ However, Q may also be true if the KB contains $R \Rightarrow Q$, in which case Q need not be retracted after all
- Systems to handle such "book keeping" are called
 Truth Maintenance Systems (TMS)

Internet shopping world

- An agent that understands and acts in an internet shopping environment
- The task is to shop for a product on the Web, given the user's product description
- The product description may be precise, in which case the agent should find the best price
- In other cases the description is only partial, and the agent has to compare products
- The shopping agent depends on having product knowledge, incl. category hierarchies

PEAS specification of shopping agent

- Performance goal
 - √ Recommend product(s) to match user's description
- <u>Environment</u>
 - √ All of the Web
- Actions
 - √ Follow links
 - √ Retrieve page contents
- Sensors
 - √ Web pages: HTML, XML

Outline of shopping agent behavior

- Start at home page of known web store(s)
 - ✓ Must have knowledge of relevant web addresses, such as <u>www.amazon.com</u> etc.
- Spread out from home page, following links to relevant pages containing product offers
 - Must be able to identify page relevance, using product category ontologies, as well as parse page contents to detect product offers
- Having located one or more product offers,
 agent must compare and recommend product
 - Comparison range from simple price ranking to complex tradeoffs in several dimensions

Summary

- An ontology is an encoding of vocabulary and relationships. Special-purpose ontologies can be effective within limited domains
- A general-purpose ontology needs to cover a wide variety of knowledge, and is based on categories and an event calculus
- It covers structured objects, time and space, change, processes, substances, and beliefs
- The general ontology can support agent reasoning in a wide variety of domains, including the Internet shopping world