Knowledge Representation
Semantic networks
Frames
<table>
<thead>
<tr>
<th>Lecture overview</th>
</tr>
</thead>
<tbody>
<tr>
<td>- Knowledge representation</td>
</tr>
<tr>
<td>- Roles and requirements of a knowledge representation</td>
</tr>
<tr>
<td>- Semantic networks</td>
</tr>
<tr>
<td>- Frames</td>
</tr>
<tr>
<td>- Inference</td>
</tr>
</tbody>
</table>
What is knowledge representation?

- What is representation?
  - Representation refers to a symbol or thing which represents (‘refers to’, ’stands for’) something else.

- When do we need to represent?
  - We need to represent a thing in the natural world when we don’t have, for some reason, the possibility to use the original ’thing’.

- The object of knowledge representation is to express the problem in computer-understandable form.
Aspects of KR

- **Syntactic**
  - Possible (allowed) constructions
  - Each individual representation is often called a sentence.
  - For example: color(my_car, red), my_car(red), red(my_car), etc.

- **Semantic**
  - What does the representation mean (maps the sentences to the world)
  - For example:
    - color(my_car, red) → ??
    - ‘my car is red’, ‘paint my car red’, etc.

- **Inferential**
  - The interpreter
  - Decides what kind of conclusions can be drawn
  - For example: Modus ponens (P, P→Q, therefore Q)
Well-defined syntax/semantics

- Knowledge representation languages should have precise syntax and semantics.
- You must know exactly what an expression means in terms of objects in the real world.
## Declarative vs. procedural knowledge

- **Declarative knowledge (facts about the world)**
  - A set of declarations or statements.
  - All facts stated in a knowledge base fall into this category of knowledge.
  - In a sense, declarative knowledge tells us what a problem (or problem domain) is all about
    - Example:
      - `cheaper(coca_cola, pepsi)`
      - `tastier(coca_cola, pepsi)`
      - `if (cheaper(x,y) && (tastier(x,y) ) → buy(x)`

- **Procedural knowledge (how something is done)**
  - Something that is not stated but which provides a mean of dynamically (usually at run-time) arriving at new facts.
  - Example: shopping script
Types of knowledge

- Domain knowledge
  - What we reason about
  - Structural knowledge
    - Organization of concepts
  - Relational knowledge
    - How concepts relate

- Strategic knowledge
  - How we reason
  - At representation level, rather than at implementation level
    - (e.g. at implementation level – control knowledge, for resolving conflicting situations)
More philosophical aspects of KR

- “What is a Knowledge Representation?” (Davis, Shrobe & Szolovits) AI Magazine, 14(1):17-33, 1993


- This paper examines the notion of KR in a deeper way, and not merely from the point of view of appropriate data structures.
What is a Knowledge Representation?

- It defines the five roles the knowledge representation plays
  - Role I: A KR is a Surrogate
  - Role II: A KR is a Set of Ontological Commitments
  - Role III: A KR is a Fragmentary Theory of Intelligent Reasoning
  - Role IV: A KR is a Medium for Efficient Computation
  - Role V: A KR is a Medium of Human Expression

- Each role defines characteristics a KR should have

- These roles provide a framework for comparison and evaluating knowledge representations
Role I: A KR is a Surrogate

- A KR is used to model objects in the world.
  - A KR is a substitute for direct interaction with the world.
  - Since a KR cannot possibly represent everything in the world, a KR must necessarily focus on certain objects and properties while ignoring others.
  - As a result only objects and properties that are relevant to reasoning are modeled.

- Consequences
  - Since our representation is not perfect, we will have errors (at least by omission) and we may even introduce new artifacts which not present
  - All sufficiently broad-based reasoning will eventually reach conclusions that are incorrect; sound inference or better KR does not help
Role I: A KR is a Surrogate

- The only complete accurate representation of an object is the object itself.

- All other representations are inaccurate.
Role II: A KR is a set of Ontological Commitments

- All representations are approximations to reality and they are invariably imperfect. Therefore we need to focus on only some parts of the world, and ignore the others.

- Ontological commitments determine what part of the world we need to look at, and how to view it.
Role II: A KR is a set of Ontological Commitments

- The ontological commitments are accumulated in layers:
  - **First layer** – representation technologies.
    For example: logic or semantic networks (entities and relations) vs. frames (prototypes)
  - **Second layer** – how will we model the world.
    Example from a frame-based system:
    “The KB underlying INTERNIST system is composed of two basic types of elements: disease entities and manifestations […] It also contains a hierarchy of disease categories organised primarily around the concept of organ systems having at the top level such categories as 'liver disease', 'kidney disease', etc”
    Commits to model prototypical diseases which will be organised in a taxonomy by organ failure
  - **Third layer (conceptual)** – which objects will be modelled.
    What is considered a disease (abnormal state requiring cure), e.g. alcoholism, chronic fatigue syndrome?
Role III: A KR is a Fragmentary Theory of Intelligent Reasoning

- Intelligent reasoning
  - “What is intelligent reasoning?”
  - The views of intelligence normally come from fields outside of AI: mathematics, psychology, biology, statistics and economics.

- Fragmentary
  - the representation typically incorporates only part of the insight or belief that motivated it
  - that insight or belief is in turn only a part of the complex and multi-faceted phenomenon of intelligent reasoning

- There are three components:
  - What does it mean to reason intelligently?
  - What can we infer from what we know?
  - What ought we to infer from what we know?
Role IV: A KR is a medium for efficient computation

- The knowledge representation should make recommended inferences efficient.
- The information should be organized in such a way to facilitate making those inferences.

- Usually a trade-off between
  - the power of expression (how much can be expressed and reasoned about in a language) and
  - how computationally efficient the language is
Role V: A KR is a medium of human expression

- A representation is a language in which we communicate
  - How well does the representation function as a medium of expression?
  - How general is it?
  - How precise?
  - Does it provide expressive adequacy?

- How well does it function as a medium of communication?
  - How easy is it for us to ‘talk’ or think in that language?
Consequences of this KR

- The spirit should be indulged, not overcome
  - KRs should be used only in ways that they are intended to be used, that is the source of their power.

- Representation and reasoning are intertwined
  - i.e. a recommended method of inference is needed to make sense of a set of facts.

- Some researchers claim equivalence between KRs, i.e. “frames are just a new syntax for first-order logic”. However, such claims ignore the important ontological commitments and computational properties of a representation.

- All five roles of a KR matter
Requirements for KR languages

- Representation adequacy
  - should to allow for representing all the required knowledge

- Inferential adequacy
  - should allow inferring new knowledge

- Inferential efficiency
  - inferences should be efficient

- Clear syntax and semantics
  - unambiguous and well-defined syntax and semantics

- Naturalness
  - easy to read and use
<table>
<thead>
<tr>
<th>Knowledge representation formalisms</th>
</tr>
</thead>
<tbody>
<tr>
<td>Production systems, expert systems</td>
</tr>
<tr>
<td>Semantic networks</td>
</tr>
<tr>
<td>Frames</td>
</tr>
<tr>
<td>Case-based reasoning</td>
</tr>
<tr>
<td>Biologically inspired approaches</td>
</tr>
<tr>
<td>- neural networks</td>
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<td>- genetic algorithms</td>
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</table>
Network notations are almost as old as logic

- Porphyry (3rd century AD) – tree-based hierarchies to describe Aristotle’s categories
- Frege (1879) - concept writing, a tree notation for the first complete version of first-order logic
- Charles Peirce (1897) – existential graphs

First implementation of semantic networks

- Masterman (1961) – defines 100 concept types which are used to define 15,000 concepts organised in a lattice
- Quillian (1967) – word concepts, defines English words in terms of other words
Semantic networks - history

- Developed by Ross Quillian, 1968, as “a psychological model of associative memory”.
- Associationist theories define the meaning of an object in terms of a network of associations with other objects in a domain or a knowledge base.
- Quillian’s model of a semantic network is based, not only on this idea of networks of information and knowledge, but on evidence that we also organise that knowledge hierarchically.
The structure of their networks was devised from laboratory testing of human response times to questions.

Questions
- “Is a canary a bird?”, “Can a canary sing?”, or “Can a canary fly?”

Response times
- “Can a canary fly?” longer response than “Can a canary sing?”.

The results of this test were analysed and Quillian concluded that humans store information at its most abstract level.
Semantic networks - history

**Figure 5.8**: Hypothetical memory structure for a three-level hierarchy. (Adapted from Collins and Quillian, 1969. Reprinted by permission of Academic Press.)
Semantic networks

- Instead of trying to recall that a canary flies, and a robin flies, and a swallow flies etc., humans remember that canaries, robins, swallows etc. are birds and that birds usually have an associated property of flying.

- More general properties, such as eating, breathing, moving etc. are stored at an even higher level associated with the concept animal.

- Thus reaction times to questions such as “Can a canary breathe?” were even longer again as more travel up the hierarchy is necessary to determine the answer.

- The fasted recall was for traits specific to the bird such as “Is a canary yellow?” or “Can a canary sing?”
Handling exceptions

- Handling exception cases also appears to be done at the most specific level.

- “Can an ostrich fly?” – quicker response than for canary
  - The conclusion reached by this is that the hierarchy ostrich -> bird -> animal is not traversed in order to understand this exception information.

- Inheritance systems allow us to store information at the highest level of abstraction – allows us to
  - reduces the size of the knowledge base
  - helps prevent update inconsistencies
What is a semantic network?

A semantic network is a type of data structure that represents knowledge as a network of concepts and connections. Each concept is represented by a node, and the connections between concepts are represented by edges. The edges can have labels that indicate the type of relationship between the concepts.

In the example shown, the concept of an animal is connected to the concepts of reptile and mammal, indicating that reptiles and mammals are types of animals. The concept of mammal has a subconcept of head, indicating that a mammal has a head. The concept of elephant is connected to the concepts of large and grey, indicating that elephants are large and grey in color. Additionally, there is a connection from elephant to apples, indicating that elephant likes apples.

Other connections include:
- Clyde is an instance of elephant.
- Nellie is an instance of elephant.
- Clyde and Nellie like apples.

These connections are represented by various edges labeled with properties such as isa, has_part, size, colour, instance_of, and likes.
Semantic networks: syntax

- Represented as a graph.
- Nodes represent concepts, actions or objects in the world.
- Links represent directional and labeled relationships between the nodes.
  - Inheritance-oriented links: "isa", "instance_of"
  - General links: "has_part", "causes"
  - Domain-specific links: "has_disease", "father_of"
Semantic network – example 1

Bilbo \(\text{instance}_{of}\) hobit \(\text{isa}\) person

find \(\text{object}\) magicRing \(\text{instance}_{of}\) ring

cave \(\text{location}\)

Gollum \(\text{owner}\)

agent

object

location

owner
Problems with semantic networks

- **Naming standards**
  - lack of naming standards for relationships
  - naming of nodes, e.g. if a node is labelled ‘chair’ does it represent:
    - A specific chair,
    - The class of all chairs,
    - The concept of a chair,
    - The person who is the chair of a meeting?

- For a semantic network to represent definitive knowledge, i.e. knowledge that can be defined, the relationship and node names must be rigorously standardised.

- Often it is hard to distinguish between
  - statements about object relationships with the world, and
  - properties of the object
Problems with semantic networks

- Searching
  - possible combinatorial explosion, especially if the response to a query is negative.
  - Semantic networks, as we know, were originally devised as models of human associative memory in which one node has links to others and information retrieval occurs due to a spreading activation of the nodes.
  - However, other forms of reasoning must be available to the brain, as it does not take a long time to answer the question “Is there a football team on Pluto?”

- Unable to represent
  - Negation
  - Quantification
  - Disjunction
Semantic networks nowadays

- Conceptual graphs (John Sowa)
- SNePS (Stuart Shapiro)
Frames

- Frames are a variant of semantic networks
- A frame is
  
  “a data structure for representing a stereotypical situation like …going to a child’s birthday party” (Minksy 1981).

```java
class Book {
    Person author;
    String title;
    int price;
}
```

- All the information relevant to a concept is stored in a single complex entity called a frame.
- Superficially, frames look like record data structures or class, however, at the very least frames also support inheritance.
In frame-based systems we refer to
- objects – Mammal, Elephant, and Nellie;
- slots – properties such as colour and size;
- slot-values – values stored in the slots, e.g. grey and large.

Slots and the corresponding slot-values are inherited through the class hierarchy
The previous example can be also represented as frames:

**Mammal:**
- subclass: Animal
- has-part: head

**Elephant:**
- subclass: Mammal
- colour: grey
- size: large

**Nellie:**
- instance: Elephant
- likes: apples

**Clyde:**
- instance: Elephant
Inheritance and default values

- In general children classes inherit the properties of the parent class
- Default values - properties that are typical for a class
- Instances or subclasses whose properties differ from these default values are able to override them.

- There are various ways of achieving overriding, for example:
  - any default value may be overridden
  - only marked slots allow the default value to be overridden
In this example only slots marked with an asterisk may be overridden.

<table>
<thead>
<tr>
<th>Mammal:</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>subclass: Animal</td>
<td></td>
</tr>
<tr>
<td>has-part: head</td>
<td></td>
</tr>
<tr>
<td>warm-blooded: yes</td>
<td></td>
</tr>
<tr>
<td>*furry: yes</td>
<td></td>
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</tbody>
</table>

<table>
<thead>
<tr>
<th>Elephant:</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>subclass: Mammal</td>
<td></td>
</tr>
<tr>
<td>*colour: grey</td>
<td></td>
</tr>
<tr>
<td>*size: large</td>
<td></td>
</tr>
<tr>
<td>*furry: no</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Nellie:</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>instance: Elephant</td>
<td></td>
</tr>
<tr>
<td>likes: apples</td>
<td></td>
</tr>
<tr>
<td>owner: Fred</td>
<td></td>
</tr>
<tr>
<td>colour: pink</td>
<td></td>
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<table>
<thead>
<tr>
<th>Clyde:</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>instance: Elephant</td>
<td></td>
</tr>
<tr>
<td>size: small</td>
<td></td>
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</table>
Inheritance

- Children classes inherit the default properties of their parent classes unless they have an individual property value that conflicts with the inherited one.

- We use the term simple (or single) inheritance when each object and class has only a single parent class.

- Multiple Inheritance considers those cases where there is more than one parent class (e.g. Clyde is an instance of both Elephant and Circus-Animal)
  - The frame system must be able to decide on precedence of inheritance

- The complication occurs when some property may be inherited from more than one parent class. Some kind of mechanism is required to select which class the property is to be inherited from.
  - The simplest solution is to define an order of precedence for the parent classes.
Let us consider the Elephant example where we consider a Circus-Animal:

<table>
<thead>
<tr>
<th>Elephant:</th>
<th>Cylde:</th>
</tr>
</thead>
<tbody>
<tr>
<td>subclass: Mammal</td>
<td>instance: Circus-Animal Elephant</td>
</tr>
<tr>
<td>has-trunk: yes</td>
<td>colour: pink</td>
</tr>
<tr>
<td>*colour: grey</td>
<td>owner: Fred</td>
</tr>
<tr>
<td>*size: large</td>
<td></td>
</tr>
<tr>
<td>*habitat: jungle</td>
<td>Nellie:</td>
</tr>
<tr>
<td>Circus-Animal:</td>
<td>instance: Circus-Animal</td>
</tr>
<tr>
<td>subclass: Animal</td>
<td></td>
</tr>
<tr>
<td>habitat: tent</td>
<td></td>
</tr>
<tr>
<td>skills: balancing-on-ball</td>
<td></td>
</tr>
<tr>
<td>*size: small</td>
<td></td>
</tr>
</tbody>
</table>

What about the size of Clyde? How can we inherit ‘habitat’ from Circus-Animal but ‘size’ from Elephant?
Both slot values and slots may be frames.

In the multiple inheritance example, we represented that Fred was Clyde’s owner. We may want to know some details about Fred, so we can use a frame to describe Fred’s properties.

Allowing a slot to be a frame means that we can specify a range of properties for a slot.

- For example, we could specify that the slot owner
  - could only take the values of the class Person,
  - has an inverse slot owns, and
  - can take multiple values (i.e. a person can own many things).
Many systems allow slots to include procedures. The term, procedural attachment is used to represent this.

A piece of program code may be placed in a slot and be run every time a value for that slot is needed. The code may also be run when a value is entered into the slot (event driven code). This code could do consistency checks or be used to propagate results to other slots.

The use of procedures and multiple inheritance can make it hard to predict what will be inferred about a given object just by looking at the set of frames. We need to know about the underlying system (the inference engine).

We say that the system has a procedural, rather than a declarative semantics, as the precise meaning of the frames depends on how the system infers new knowledge.
Inference in semantic networks and frames

- Semantic networks and frames provide a fairly simple and clear way of representing the properties and categories of objects.
- A basic type of inference is defined whereby objects may inherit properties of parent objects.
- However, inheriting properties from more than one parent, or defining conflicting properties if often problematic.
Software - Protégé

http://protege.stanford.edu/
Conclusions

- Knowledge representation
- Roles and requirements of a knowledge representation
- Semantic networks
- Frames
- Inference
Exercise (1)

- Represent the following as a set of frames

  - The aorta is a particular kind of artery which has a diameter of 2.5cm. An artery is a kind of blood vessel. An artery always has a muscular wall, and generally has a diameter of 0.4cm. A vein is a kind of blood vessel, but has a fibrous wall. Blood vessels all have tubular form and contain blood.
Exercise 1 (solution)

Blood vessel:
  form: tubular
  contain: blood

Artery:
  subclass: Blood vessel
  wall_type: muscular
  diameter: 0.4cm

Vein:
  subclass: Blood vessel
  wall_type: fibrous

Aorta:
  subclass: Artery
  diameter: 2.5cm
Exercise (2)

- Represent the following as a semantic network

We consider any individual studying or conducting research at a university to be an academic. Within the academic community, there are two categories: students and staff.

Students get some form of funding and staff get a salary. Students who are studying for their primary degree are called undergraduates, and attend a particular course (e.g. Mathematics, Computer Science, Geography, etc.). All other students are called post-graduates and have a primary degree. They also have some research area (e.g. Artificial Intelligence).

Three categories of staff exist: lecturers, demonstrators, and researchers. Lecturers give a course (e.g. C/C++ Programming), and demonstrators provide support for those courses. On the other hand, researchers conduct research into a particular research area.

John is a student studying Computer Science. Mary is a lecturer in Computer Science.