Review 2: Logic-Based Inference

- **Deductive Inference in FOPL** (Chapter 9, Sections 10.3, 10.4)
 - Convert first order sentences to clause form
 - Definition of clauses
 - Conversion procedure
 - step 1: Eliminate implication and equivalence symbols
 - step 2: Move all negation symbols to individual predicates
 - step 3: Eliminate all existential quantifiers (Skolemization)
 - step 4: Eliminate all universally quantifiers
 - step 5: Convert the sentence to conjunctive normal form
 - step 6: use parenthesis to separate all disjunctions, then drop all v's and ^'s
 - Unification (obtain mgu θ)
 - Two terms x and y can be unified only if one of them, say x, is a variable and x does not appear anywhere in y. Then x/y is added into the substitution θ .
 - When one binding variable/term is found, apply it to the remainders of both argument lists and to previously found bindings before proceeding to unify other arguments
 - Two argument lists of the same predicate are unifiable if every corresponding pair of terms, one from each list, is unifiable
 - Resolution
 - Two clauses C1 and C2 can be resolved if one contain literal P and the other contains \sim P and the argument lists of P and \sim P can be unified with mgu θ
 - The resulting clause (resolvant) is composed of all literals of C1 and C2 except P and
 ~P, subject to variable substitution according to θ.
 - Resolution Refutation
 - Write the axioms as FOL sentences and convert them into clause form
 - Write the goal (theorem) as a FOL sentence
 - Negate the goal and convert it to clause form
 - Select a pair of clauses for resolution which are i) resolvable, and ii) promising toward deriving a null clause,
 - Inference stops when a null clause is derived
 - Control strategies
 - Depth-first: incomplete
 - Breadth-first: complete
 - Unit resolution (one of the two parent clauses is a singleton clause): incomplete in general but complete for Horn clauses
 - Input resolution (one of the two parent clauses is from the original set of clauses): incomplete in general but complete for Horn clauses
 - Set of support: complete
 - Linear resolution (one of the two parent clauses is either from the original set of clauses or is an ancestor of the other parent clause): complete
 - Horn clauses and logic programming
 - Definition of Horn clauses
 - Advantages and limitations of using Horn clauses
 - Logic programming as a general purpose programming language (viewing and resolution as function calls answer extraction)
 - Features of Prolog (Horn clauses, top-down/left-right, depth-first plus backtracking)

- Advantages
 - Clearly defined semantics (least ambiguous
 - Expressiveness
 - Natural for some domains
- Disadvantages
 - Semantics is too rigid
 - Inefficiency (inference is NP}-hard with complete control strategies; semi-decidable)
 - Unnatural for many domains

• **Production (Rule-Based) Systems**] (Section 10.5)

- System components: WM, rule base, inference engine (rule interpreter)
- Inference procedure
 - Cycle of three phases: match, conflict-resolution, act/fire
 - Forward and backward inference
- Conflict resolution
 - conflict set
 - conflict resolution policies (refraction, specificity, recency, priority/rule-ordering)
- Advantages
 - Simplicity (for both language and inference)
 - Efficiency
 - Modularity (easy for KB maintenance)
 - Natural for many application domains
- Disadvantages
 - No clearly defined semantics (based on informal understanding)
 - Incomplete inference procedure
 - Unpredictable side effects of ordering of rule applications
 - Less expressive (may not be suitable for some applications)

• Structured representation

- Semantics (associative) networks
 - labeled nodes: objects, classes, concepts
 - Labeled directed links: relations (associations) between nodes
 - Reasoning about associations (marker passing and spreading activation)
- ISA hierarchy and property inheritance
 - Super/subclass and instance/class relation
 - Inference by inheritance
 - Multiple inheritance (from different parents, from ancestors of different distances)
 - Exceptions in inheritance
- Frame Systems
 - Definition (stereotypical views of the world; record like structure)
 - Slots, their values and facets
 - Procedural attachment and how they work (if-added, if-needed, if-updated)
 - Frames from different perspectives

Default reasoning

- Definition (inference is drawn in the absence of info to the contrary) and examples

- Default reasoning is non-monotonic, and it totally undecidable
- How production systems and semantic networks (and frame systems) handle simple default reasoning

Review 3: Abduction, Uncertainty, and Probabilistic Reasoning

Abduction

- Definition
- Difference between abduction, deduction, and induction
- Characteristics of abductive inference
 - Inference results are hypotheses, not theorems (may be false)
 - There may be multiple plausible hypotheses
 - Reasoning is often a hypothesize-and-test cycle
 - Reasoning is non-monotonic
 - Sources of uncertainty (uncertain data, knowledge, and inference)

• Simple Bayesian approach to evidential/diagnostic reasoning

- Bayes' theorem
- Conditional independence (and evidence) and single fault assumptions
- Methods for computing posterior probability and relative likelihood of a hypothesis, given some evidence

$$\begin{split} &P(H_i \mid E_1, ..., E_l) = \frac{P(E_1, ...E_l \mid H_i)P(H_i)}{P(E_1, ...E_l)} \\ &rel(H_i \mid E_1, ..., E_l) = P(E_1, ..., E_l \mid H_i)P(H_i) = P(H_i)\prod_{j=1}^{l} P(E_j \mid H_i) \\ &P(H_i \mid E_1, ..., E_l) = \frac{rel(H_i \mid E_1, ..., E_l)}{\sum\limits_{k=1}^{n} rel(H_k \mid E_1, ..., E_l)} = \frac{P(H_i)\prod_{j=1}^{l} P(E_j \mid H_i)}{\sum\limits_{k=1}^{n} P(H_k \mid E_l, ..., E_l)} \end{split}$$

Evidence accumulation

$$\begin{split} rel(H_i \mid E_1, ..., E_l, \sim E_{l+1}) &= (1 - P(E_{l+1} \mid H_i)) rel(H_i \mid E_1, ..., E_l) \\ rel(H_i \mid E_1, ..., E_l, E_{l+1}) &= P(E_{l+1} \mid H_i) rel(H_i \mid E_1, ..., E_l) \end{split}$$

- Limitation
 - Assumptions unreasonable for many problems
 - Not suitable for multi-fault problems
 - Can not represent causal chaining

• Bayesian belief networks (BBN)

- Integration of probability theory and causal networks
- Definition of BBN (DAG and CPD).

$$P(x_i | p_i)$$
 where p_i is the set of all parent nodes of x_i

Independence assumption

$$P(x_i | p_i, q) = P(x_i | p_i)$$

Computing joint probability distribution

$$P(x_1,...,x_n) = \prod_{i=1}^n P(x_i | p_i)$$

- Inference (e.g., belief update, MAP) is NP}-complete (exponential time)
- BBN of noise-or gate (advantages and limitations)
- Learning BBN from case data (difficulty in learning the DAG)

• Dempster-Shafer theory (for representing ignorance)

- Difference between probability of an event and ignorance
- How to represent uncommitted belief (ignorance)
 - Lattice of subsets of frame of discernment
 - Basic probability assignment (function m(S))
 - Bel(S), Pls(S), and belief interval
- Problem with this theory (high complexity)

• Fuzzy set theory (for representing vague linguistic terms)

- Difference between fuzzy sets and ordinary sets
- Fuzzy membership functions
- Rules for fuzzy logic connectives
- Problems with fuzzy logic (comparing with probability theory)

• Uncertainty in rule-based system (certainty factors in MYCIN)

- CF of WM elements
- CF of rules
- CF propagation
- Problems with CF