

Exam 2 Review

Knowledge Representation

- **Production (Rule-Based) Systems**
 - 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**
 - Semantic (associative) networks
 - Labeled nodes: objects, classes, concepts
 - Labeled directed links: relations (associations) between nodes
 - reification
 - 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/default reasoning
 - Frame Systems
 - Definition (stereotypical views of the world; record like structure)
 - Slots, their values and facets
- **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 rule-based systems and semantic networks (and frame systems) deal with simple default 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
 - Inherently uncertain

Planning

- **Situation calculus planning**
 - Reasoning about change in the world
 - Representing states and state changes by actions
 - Planning by theorem proving (expensive)
- **STRIPS planning**
 - State, goal: using ground literals
 - Actions/operators: add-list and delete-list
 - Simple STRIP planning (assuming goals are independent)
 - Limitations (Sussman's anomaly) because subgoals are satisfied independently
- **Partial order planner (POP)**
 - Difference between total order (linear) and partial order (non-linear) planning
 - Least commitment principle
 - Causal links and ordering constraints
 - A complete POP
 - Linearizing a partial plan

Uncertainty and Probabilistic Reasoning

- Simple Bayesian approach to evidential/diagnostic reasoning
 - Bayes' theorem
 - Conditional independence and single fault assumptions
 - Computing posterior probability and relative likelihood of a hypothesis, given some evidence
 - Limitation
 - Assumptions unreasonable for many problems
 - Not suitable for multi-fault problems
 - Can not represent causal chaining
- **Bayesian networks (BN)**
 - Definition of BN (DAG and CPT).
 $P(x_i | \pi_i)$ where π_i is the set of all parent nodes of x_i
 - **Conditional independence assumption**
 - $P(x_i | \pi_i, q) = P(x_i | \pi_i)$
 - d-separation
 - Markov blanket

- **Computing joint probability distribution from CPT: chain rule**
- Inference
 - NP-hard
 - Exact methods (**enumeration**, ideas of variable elimination, junction tree and belief propagation)
 - Approximate methods (stochastic sampling, MCMC, loopy propagation)
- BN of noise-or gate (advantages and limitations)
- Learning BN from case data (difficulty in learning the DAG)
- **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)
- **Decision making under uncertainty**
 - Actions, uncertain outcomes, and utility
 - **Expected utility**
 - **Maximum expected utility (MEU) principle**

$$EU(\alpha | E) = \max_A \sum_i U(\text{Result}_i(A)) p(\text{Result}_i(A) | E, \text{Do}(A))$$
 - Decision network (influence diagram)
 - Chance nodes, decision nodes, and utility nodes
 - Value of perfect information (VPI): definition, meaning, how to compute

$$\text{VPI}(X) = (\sum_k p(x_k | E) (EU(\alpha_{xk} | x_k, E)) - EU(\alpha | x_k, E))$$

Learning

- **Supervised, unsupervised, and reinforcement learning**
- **Decision tree learning**
 - Decision tree (nodes and arcs)
 - **Information gain** (definition and how to use it to construct a decision tree)

$$\text{Info}(T) = I(P) = - \sum_i p_i \log(p_i) ; \text{Info}(X, T) = \sum |T_i|/|T| * \text{Info}(T_i)$$
 - **Overfitting problem and cross-validation**
 - Generating rules from decision tree
 - Limitations of decision tree learning
- **Neural Networks**
 - Comparisons between Von Neumann machine and human brain and artificial neural networks
 - Perceptron: the network, the learning rule, and the limitation (**linear separable problems**)
 - Feed forward networks: hidden nodes of non-linear functions, learning rule (**gradient descent**), what does error back propagation mean?
 - Advantages and limitations
- **Support vector machine (SVM)**
 - Basic ideas
 - **Maximum margin** classifier to increase its robustness and generalization power

- What are support vectors
- Maximum margin can be computed by quadratic programming (QP)
- Overcome linear separability problem by converting the problem into a higher dimension space
 - converting the problem into a higher dimension feature space
 - **kernel functions** help the dimensionality explosion in QP

Note: materials covered in the class but not listed in this document will not be tested in Exam 2.