

AI topics/questions

1. Intro
 - a. The four approaches to AI.
 - b. The Turing test
 - c. Acting rationally. The rational agent.
2. Agents
 - a. Agent function, agent program, agent types/architectures.
 - b. Environment properties: Observable, deterministic, static, single-agent.
 - c. The reflex agent architecture.
 - d. The utility-based agent architecture.
3. Problem-solving with search
 - a. Problem types.
 - b. The single-state problem formalization.
 - c. The general tree search algorithm (pseudo-code only).
 - d. States vs. nodes
 - e. The four evaluation metric/properties for search strategies: completeness, space-complexity, time-complexity, optimality (branching factor, diameter of the state space).
 - f. Uninformed search.
 - i. Breadth-first (concept, pseudocode, properties)
 - ii. Depth-first (concept, pseudocode, properties)
 - iii. Iterative deepening (depth-limited depth-first) search (concept, pseudocode, properties)
 - iv. Comparison of properties.
4. Informed search
 - i. Heuristic function
 1. admissibility
 2. dominance
 3. derivation with relaxing the problem
 - ii. Greedy search (concept, pseudocode, properties)
 - iii. A* (concept, pseudocode, properties) optimality with informal proof
5. Local search
 - a. The hill-climbing algorithm (pseudocode)
 - b. Problems with the hill-climbing algorithm
 - c. Simulated annealing
6. Logic
 - a. The concept of general purpose inference and domain specific knowledge-base.
 - b. Logic: syntax and semantics (conceptualization).
 - c. The syntax of propositional logic.
 - d. The concept of models wrt KBs and the model-based definition of semantic inference: entailment.
 - e. Implementation of entailment in propositional logic (pseudocode).
 - f. (Syntactic) inference: elementary steps: modus ponens, resolution.

- g. Relation between entailment and (syntactic) inference: soundness, completeness.
 - h. The forward-chaining proof method (pseudocode)
 - i. Conversion of a KB to CNF form.
 - j. The resolution-based proof method (pseudocode)
 - k. The first-order logic:
 - i. Advantages
 - ii. Quantifiers
7. Uncertainty
- a. The subjective interpretation of probability
 - b. Decision theory: the binary decision problem (which action?)
 - c. Probability theory
 - i. Atomic events, composite events, joint distribution
 - ii. Conditional probability, the chain rule
 - iii. The Bayes rule,
 - 1. prior and posterior probabilities
 - 2. relevance: causal and diagnostic direction
 - iv. Independence, conditional independence
 - d. Inference by enumeration
 - e. The naive Bayes model.
 - i. The product form for the joint distribution.
 - ii. Diagnostic inference
 - iii. The structure.
8. The Bayesian networks.
- a. Syntax.
 - b. Compactness (for binary random variables with max k parents).
 - c. Global semantics (the product decomposition of the joint wrt the structure)
 - d. Construction steps.
9. Inference in Bayesian networks.
- a. Tasks: simple query, composite query, relevance
 - b. Inference by enumeration (pseudocode).
 - c. Inference by stochastic simulation
 - i. Sampling from an empty network (concept, pseudocode).
 - ii. Gibbs sampling, Markov blanket.
10. Temporal probability models
- a. Definition of a Markov process (homogeneous).
 - b. Definition of a Hidden Markov model (homogeneous).
 - i. Inference tasks: definitions of filtering, most likely explanation, smoothing.
 - 1. Filtering (concept, derivation, pseudocode)
11. Decision theory
- a. Utility theory, preferences, the conditions for the existence of a utility function.
 - b. The maximum expected utility principle.
 - c. Decision network: elements and structure.
 - d. Value of perfect information, formula

12. Learning

- a. The function approximation view of inductive learning.
- b. The Ockham principle, Hume's problem of induction.
- c. Bayesian learning
 - i. Bayes rule
 - ii. Posterior probability of a model/hypothesis
 - iii. Prediction using averaging, MAP and ML approximations.
- d. The learning curve.
- e. The bias-variance dilemma
- f. Probably Approximately Correct (PAC) learning
 - i. definition
 - ii. the misclassification rate as loss
 - iii. derivation of sample complexity of concept learning in i.i.d. context (independent identically distributed)
- g. Reinforcement learning

13. The decision tree representation.

- a. Expressivity
- b. Cardinality
- c. Learning method