

# ***Applications of Artificial Intelligence***

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## **Chapter 4: Knowledge-Based Systems**

### **4.5: Concluding Comparison of the Different Reasoning Techniques**

# Application from practice: Technical diagnosis

## Run time system:

(knowledge-based systems call this **problem solver / inference engine**)

### Input:

- Setting certain control inputs
- Observing values depending on this setting

### Output:

- A unique instruction how to repair which component

***This is where diagnostic systems do **not** differ !***

# Application from practice: Technical diagnosis

## Knowledge-based diagnosis:

### 1) Knowledge acquisition: Input into knowledge base

- symptom-based (rule-based)
  - model-based
  - case-based (machine learning)
- } as alternatives

### 2) Knowledge structure

- depends on knowledge acquisition

### 3) Knowledge processing by the problem solver

- depends on knowledge structure

***This is where diagnostic systems may differ !***

# 1. Symptom-Based Diagnosis

## Input to knowledge base:

- Causing and manifest faults for the overall system
- Possible symptoms (measurements)
- Relations between faults and symptoms (rules)

## Structure of knowledge base:

- Semantic network (e.g., fault networks, decision trees)

## Job of inference engine:

- Navigation in semantic network

***This is „classical“ expert system technology***

## 2. Model-Based Diagnosis

### Goal:

- fast knowledge acquisition
- exact and provable solution of problem solver

### Input to knowledge base:

- system model: hierarchical structure of the system (+ how the components are connected)
- component models

### Structure of knowledge base:

- constraint network (assembled automatically)

### Job of inference engine:

- GDE approach: conflict-based candidate generation
- sophisticated acceleration techniques in order to get reasonable run time behaviour (only discussed for candidate generation, others not discussed in class)

# 3. Case-Based Diagnosis (Machine Learning)

## Input to knowledge base (supervised approach only):

- Cases with complete symptom vector and associated faults (classified unambiguously)

## Structure of knowledge base:

### a) Classical AI, with similarity measure:

- Similarity measure for incomplete symptom vectors (often weighted between different types of symptoms)

### b) with neural networks:

- Neural network with input layer (for symptom vector) and output layer (for faults) and (optionally) intermediate layer of nodes and edges, marked by variable weights.

## Job of inference engine:

- a)**
  - For a new vector given, find the most similar symptom vector of the knowledge base.
  - Assign the same fault to the new vector as associated to the reference vector in the knowledge base (possibly with a probability value).
- b)**
  - Apply new symptom vector to the input layer of the network.
  - Read the associated fault from the output layer.

# Systematic classification of inference techniques

- **heuristic:**

**if <features> then <solution>**  
(usually the solution has got disjunctive alternatives)

- **causal:**

- overlapping classification:

**if <solution> then <features>**

- structural classification:

**local behavioural model => system function**

(search for the best behavioural models being consistent with the observed overall system behaviour)

# Systematic classification of inference techniques

- **case-based (machine learning approach):**

Given cases with **features** and **solution**

Apply regression technique (**interpolation**)

- with similarity measure:

arbitrary regression

- in neural networks

distributed linear regression

- in data mining (unsupervised approach):

**features from knowledge base => new correlations**

Supplementary, apply one of the other methods (heuristic or causal)



# Systematic classification of inference techniques

## Classification of knowledge-based inference by depth

- heuristic *for relatively flat knowledge*
- causal *for flat and deep knowledge*
- case-based (similarity measure, neural network, data mining) *for very flat knowledge*



In principle, this may be arbitrarily combined with other dimensions of knowledge quality:

- certain vs. uncertain (consider the probability of a statement)
- exact vs. fuzzy (consider the accuracy of a statement)

# Concluding comparison for applicability in practice

	rule-based	case-based	model-based
fast run time component	<b>++</b>	<b>++</b>	<b>o</b>
fast knowledge acquisition	<b>o</b>	<b>++</b>	<b>+</b>
fits to systems of complex structure	<b>--</b>	<b>++</b>	<b>++</b>
fits to systems containing complex components	<b>+</b>	<b>++</b>	<b>--</b>
reusability of knowledge	<b>o</b>	<b>--</b>	<b>++</b>
fits to diagnosis of unknown faults	<b>--</b>	<b>a) -- b) -</b>	<b>+</b>
is readily available at product launch	<b>o</b>	<b>a) -- b) -</b>	<b>++</b>
provable reliability of diagnoses	<b>+</b>	<b>a) o b) --</b>	<b>++</b>