

UNIVERSITY OF CALGARY
Computer Science

CPSC 433 - Artificial Intelligence

Final Review

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- What is AI?
- What are the problems AI addresses?
- Search vs computation

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Muddiness

CATEGORY	FACTOR	CLEAN	MUDDY
External Env. (external to brain)	Awareness	Known	Unknown
	Complexity	Simple	Complex
	Controlledness	Controlled	Uncontrolled
	Variation	Fixed	Changing
	Foreseeability	Foreseeable	Nonforeseeable
Input	Rawness	Symbolic	Real sensor
	Size	Small	Large
	Background	None	Complex
	Variation	Simple	Complex
	Occlusion	None	Severe
Internal Env. (internal to brain)	Activeness	Passive	Active
	Modality	Simple	Complex
	Multi-modality	Single	Multiple
	Size	Small	Large
	Representation	Given	Not given
Output	Observability	Observable	Unobservable
	Impossibility	Impossible	Nonimpossible
	Time coverage	Simple	Complex
	Terminalness	Low	High
	Size	Small	Large
	Modality	Simple	Complex
	Multimodality	Single	multiple

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Search Models

A:(S,T)	Set-based	And-tree-based	Or-tree-based
S:possible states	S <i>F: set of facts</i>	S_A <i>Atree: (pr, sol:{yes,?}, b, n ≥ 0)</i>	S_V <i>Otree: (pr, sol:{yes,?,no}, b, n ≥ 0)</i>
T:SxS = possible Transitions	$T: \{(s,s') \mid \exists A \rightarrow B \in \mathcal{E}xt \bullet A \subseteq s \wedge s' = (s-A) \cup B\};$ $\mathcal{E}xt \subseteq \{A \rightarrow B \mid A, B \subseteq \mathcal{F}\}$	T_A <i>Erw</i> ↓ ↓ ↓ (pr,?,b) ↓ (pr,?,b)	T_V <i>Erw</i> ↓ ↓ ↓ (pr,?,b)

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Search Processes

P:(A,Env,K)	Set-based	And-tree-based	Or-tree-based
$K: S \times Env \rightarrow S$	$\mathcal{K}(s,e) = (s-A) \cup B$		
Utility: $S \times S \times Env \rightarrow Nat$	$\forall A' \rightarrow B' \in \mathcal{E}xt \mid A' \subseteq s \bullet$ f_{Wert} f_{Wert}	f	f
Select: 2	A $\forall A' \rightarrow B' \in \mathcal{E}xt \mid A' \subseteq s \bullet$ f_{Wert} f_{Wert} f_{select}	f	f

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Instances and Goals

Ins=(s)	Set-based	And-tree-based	Or-tree-based
$G: s \rightarrow \{yes,no\}$	s no more expansions possible	$s =$ (pr', yes) v ((pr', ?, b G the sols are compatible) v no more solutions	$s =$ (pr', yes) v ((pr', ?, b G) v can't process any leaves

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Logics

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Logics

- General logics: T, P, J, Q, I
- Propositional logic
- Predicate logic
- Logic of EMycin
- Tautologies, Satisfiable, Unsatisfiable

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Prolog

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EMycin

Let h be an object-attribute-value triple and e a set of production rules. If $P_1 \wedge \dots \wedge P_n \rightarrow h$ is the only rule in e , then we get

$$MB(h, e) = I(P_1 \wedge \dots \wedge P_n \rightarrow h) * \max(0, \min(I(P_1), \dots, I(P_n)))$$

If $e = \{e_1, e_2\}$, then we get

$$MB(h, \{e_1, e_2\}) = 0, \text{ if } MD(h, \{e_1, e_2\}) = 1$$

$$MB(h, \{e_1, e_2\}) = MB(h, \{e_1\}) + MB(h, \{e_2\}) * (1 - MB(h, \{e_1\}))$$

For more elements just iterate this.

MD is computed similarly, except that e contains all production rules of the form $P_1 \wedge \dots \wedge P_n \rightarrow \neg h$

☞ application of **Bayes formula** for conditional probabilities

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Frames: XML/DTD

```

<!DOCTYPE CATALOG [
  <ENTITY AUTHOR "John Doe">
  <ENTITY COMPANY "JD Power Tools, Inc.">
  <ENTITY EMAIL "jd@jd-tools.com">
  <ELEMENT CATALOG (PRODUCT+)>
  <ELEMENT PRODUCT
    (SPECIFICATIONS+, OPTIONS?, PRICE+, NOTES?)>
  <!ELEMENT OPTIONS (#PCDATA)>
  <!ELEMENT PRICE (#PCDATA)>
  <!ELEMENT SPECIFICATIONS (#PCDATA)>
  <!ELEMENT NOTES (#PCDATA)>
  <product name="widget" partnum="T123" inventory="Backordered">
  <specifications weight="120kg" power="240v">
  <price MSRP="$50.00">
  </product>
  ]
  
```

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Semantic Nets: Conceptual Graphs

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Existential Graphs

- 1. Insertion**
Any graph can be inserted in to an oddly cut region
- 2. Erasure**
Any graph in an even number of cuts can be removed
- 3. Iteration**
Copy any graph into a region that occurs in all the original occurs in
- 4. Deiteration**
An graph obtained from iteration can be erased
- 5. Double Cut**
2 cuts around a graph can be erased, and 2 cuts can be added around a grap

Show: $\frac{P \cdot P \rightarrow Q}{Q}$

P P Q

P P Q by deiteration

P Q by double cut

Q by erasure

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Semantic Nets: Conceptual Graphs

- Build a conceptual dependency representation for the following sentences:
 - John eats a steak
 - John ate pizza yesterday
- Build the graph for the following question and match it against the knowledge base from above:
 - Who had pizza yesterday?

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Neural Nets

$x_{ij} = f_{act}(x_{i-1,1} * w_{i-1,1,j} + \dots + x_{i-1,ni-1} * w_{i-1,ni-1,j})$

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Neural Nets: Perceptron

- No hidden layers
- $f_{act}(a_1, \dots, a_m) = \text{sig}(a_1 + \dots + a_m)$
- $f_{learn}(w_{0j1,old}, d, x_{01}, \dots, x_{0k}, c) = w_{0j1,old} + c * (d - \text{sig}(\sum x_{0i} * w_{0i1})) * x_{0i}$
where d is the expected output of the web
- Usually there are more inputs to a perceptron than what is suggested by the function to learn: bias-nodes allow for more learning accuracy

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Neural Nets: Perceptron

Perform the learning for a perceptron with two input nodes and one bias node with constant value of 1 and a learning rate $c = 0.3$

Let the initial weights be
 $w_{011} = 0.3$; $w_{021} = 0.4$; $w_{031} = -0.2$

Training data:
 $(1, 1) \rightarrow 1$; $(9.4, 6.4) \rightarrow -1$; $(8, 7.7) \rightarrow -1$;
 $(0.5, 2.2) \rightarrow 1$

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Constraints: Rules of thumb

- Theoretical investigations: what is possible?
 - Logics
- Knowledge already in very similar format
 - take format
- Hierarchical structures / inheritance
 - semantic nets, frames
- Represent certain input-output behavior
 - neural nets
- Laws, rules
 - rule sets

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Learning

In general:
Structuring (or restructuring) of knowledge (due to experiences)

In AI systems:
Restructuring in order to **improve** behavior of system

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Cooperation: TECH

- TEams for Cooperative Heterogeneous Search
- Basic idea: **different agents** employ different search processes working on the **same search instance**; both different controls and different search models
- From time to time an agent selects pieces of **different kinds of knowledge** to communicate them to the other agents
- An agent receiving pieces of knowledge **filters** them according to its needs and then integrates them into its search

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Cooperation: TECH

- Positive** information to be integrated into the **search state**: (partial) solutions to problem or subproblems
- Negative** information to be integrated into the **search state**: unsolvable subproblems, partial solutions not leading to a solution
- Positive control** information: control parameters, focus pieces of solutions
- Negative control** information: transitions to avoid

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Agent Communication: CASA

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