

# DM533 (5 ECTS - 2nd Quarter) Introduction to Artificial Intelligence

Introduktion til kunstig intelligens

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# What is AI?

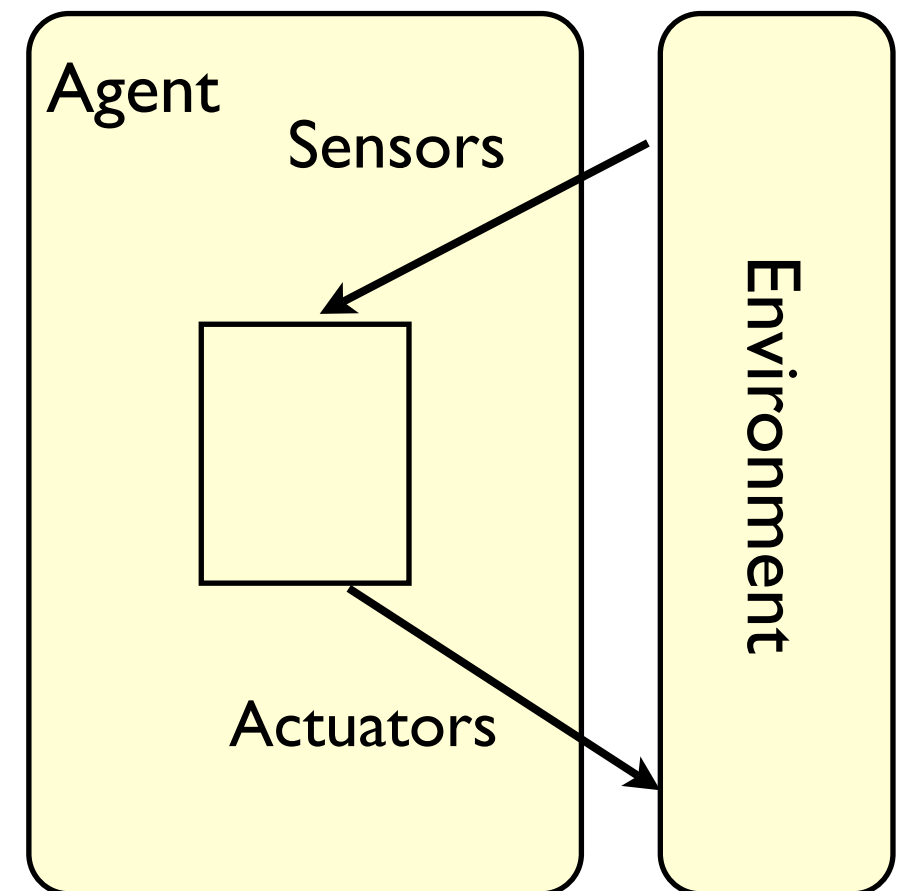
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**Agents:** something that acts, a computer program, a robot

**Rationality:** acting so as to achieve the best outcome, or when there is uncertainty, the best expected outcome

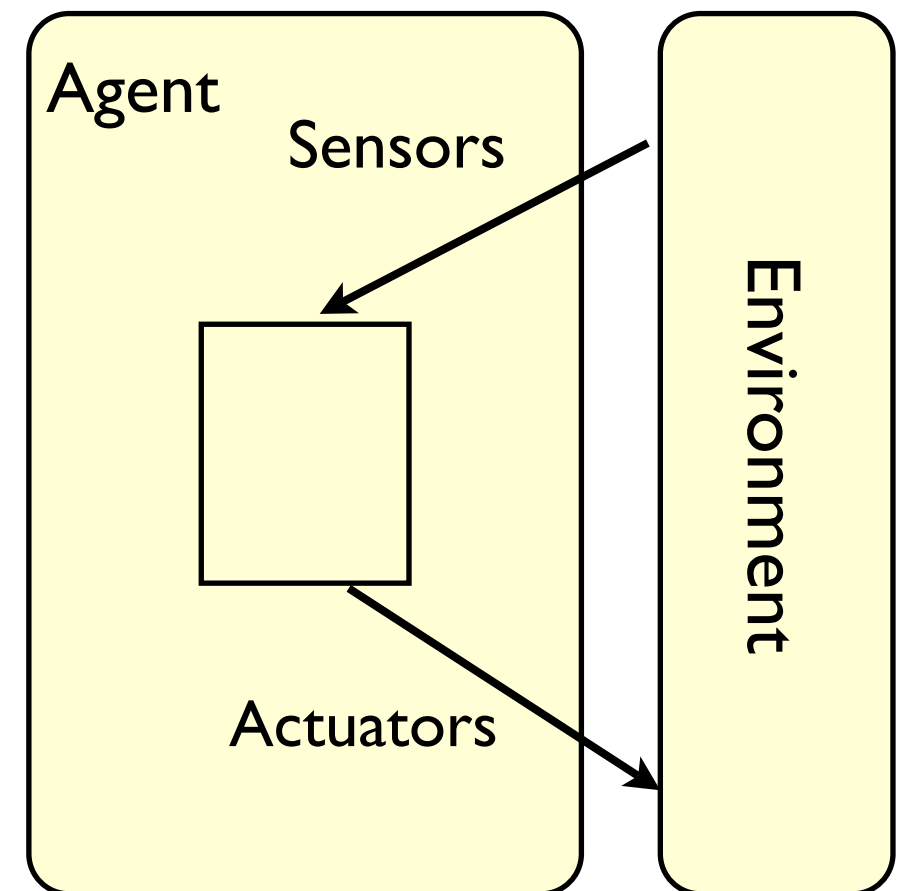


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- ➔ In complicated environments, perfect rationality is often not feasible

# History

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Today: AI is a branch of computer science with strong intersection with operations research, decision theory, logic, mathematics and statistics

# Contents

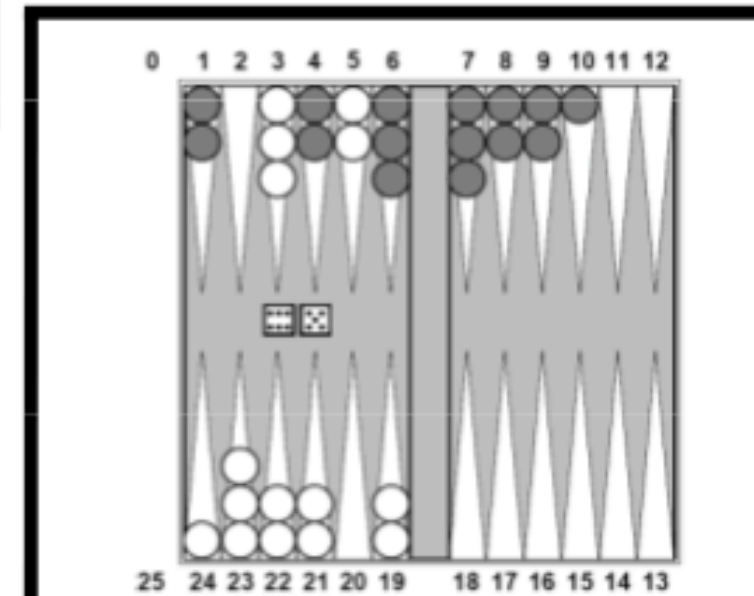
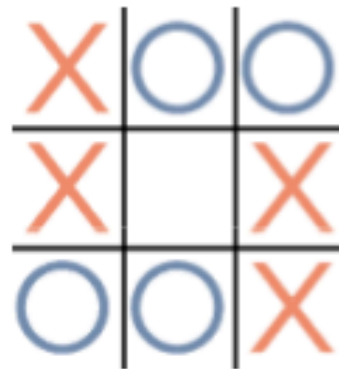
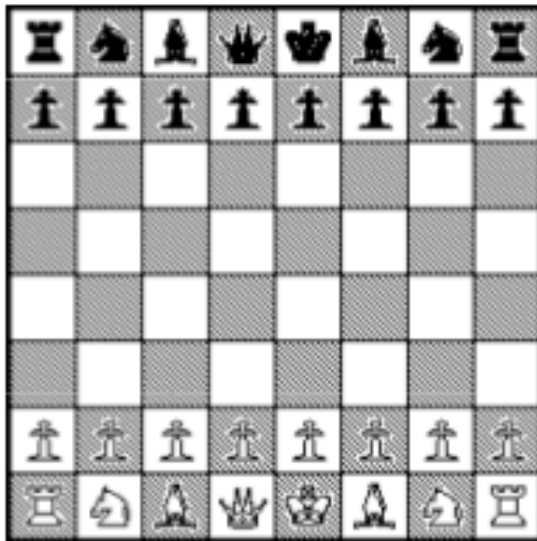
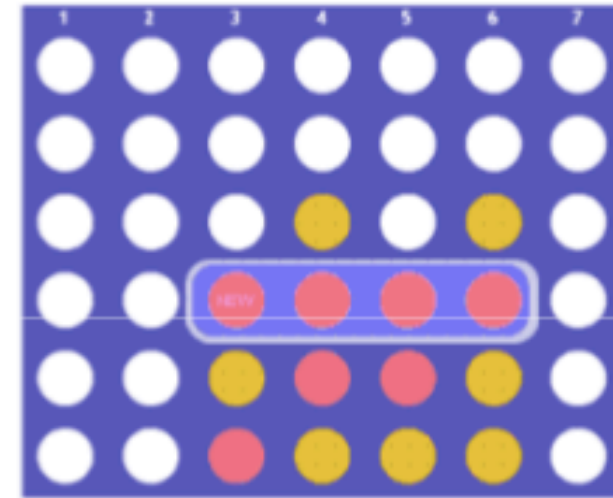
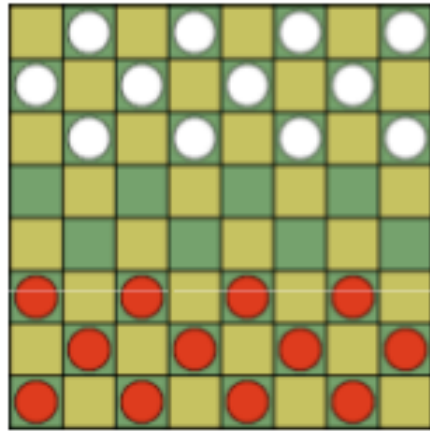
1. Introduction, Philosophical aspects (2 lectures)
2. Problem Solving by Searching (2 lectures)
  - Uninformed and Informed Search
  - Adversarial Search: Minimax algorithm, alpha-beta pruning
3. Knowledge representation and Inference (3 lectures)
  - Propositional logic, First Order Logic, Inference
  - Constraint Programming (Comet or Prolog)
4. Decision Making under Uncertainty (4 lectures)
  - Probability Theory + Utility Theory
  - Bayesian Networks, Inference in BN,
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5. Machine Learning (4 lectures)
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## 2. Problem Solving by Searching

- Uninformed and Informed Search
- **Adversarial Search**: Minimax algorithm, alpha-beta pruning

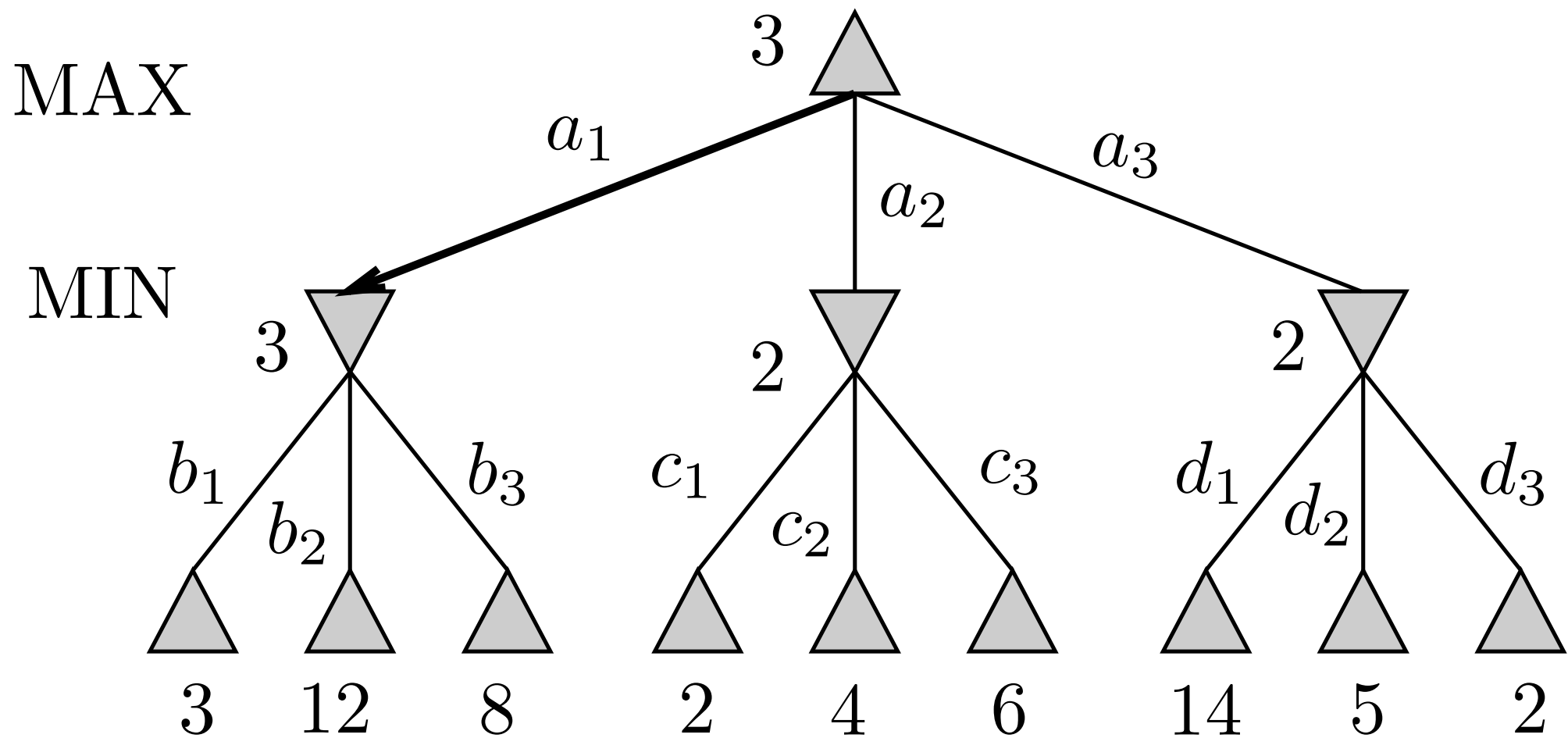
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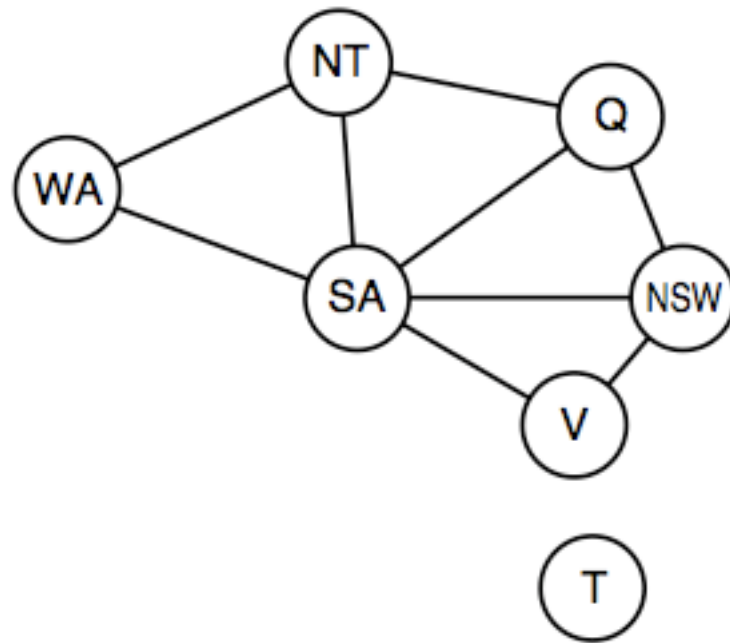


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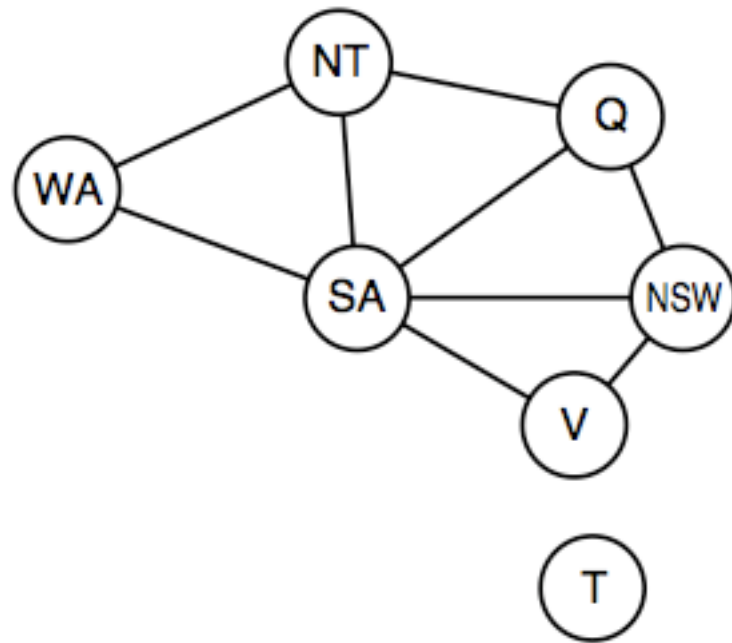
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$$\begin{aligned}
 & Diff(wa, nt) \wedge Diff(wa, sa) \wedge \\
 & Diff(nt, q) Diff(nt, sa) \wedge \\
 & Diff(q, nsw) \wedge Diff(q, sa) \wedge \\
 & Diff(nsw, v) \wedge Diff(nsw, sa) \wedge \\
 & Diff(v, sa) \Rightarrow Colorable() \\
 & Diff(Red, Blue) \quad Diff(Red, Green) \\
 & Diff(Green, Red) \quad Diff(Green, Blue) \\
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Finding a solution to the  
 Constraint Satisfaction Problem  
 corresponds to infer coloring in FOL

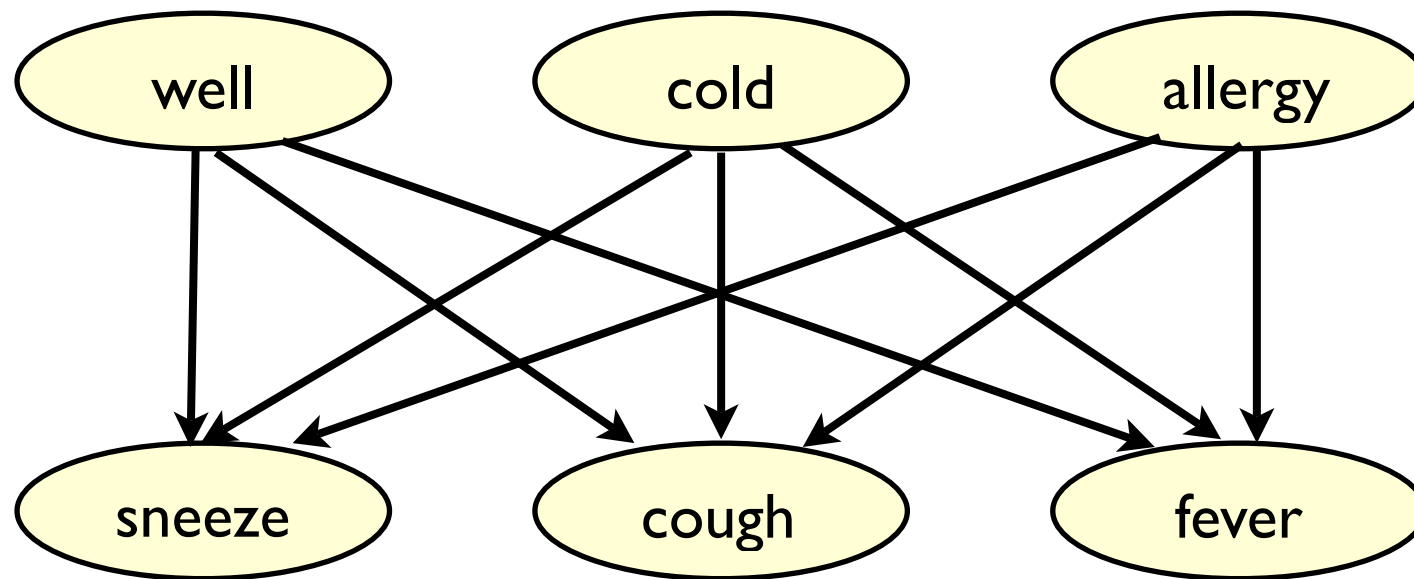


# 4. Decision Making under Uncertainty

- Probability Theory + Utility Theory
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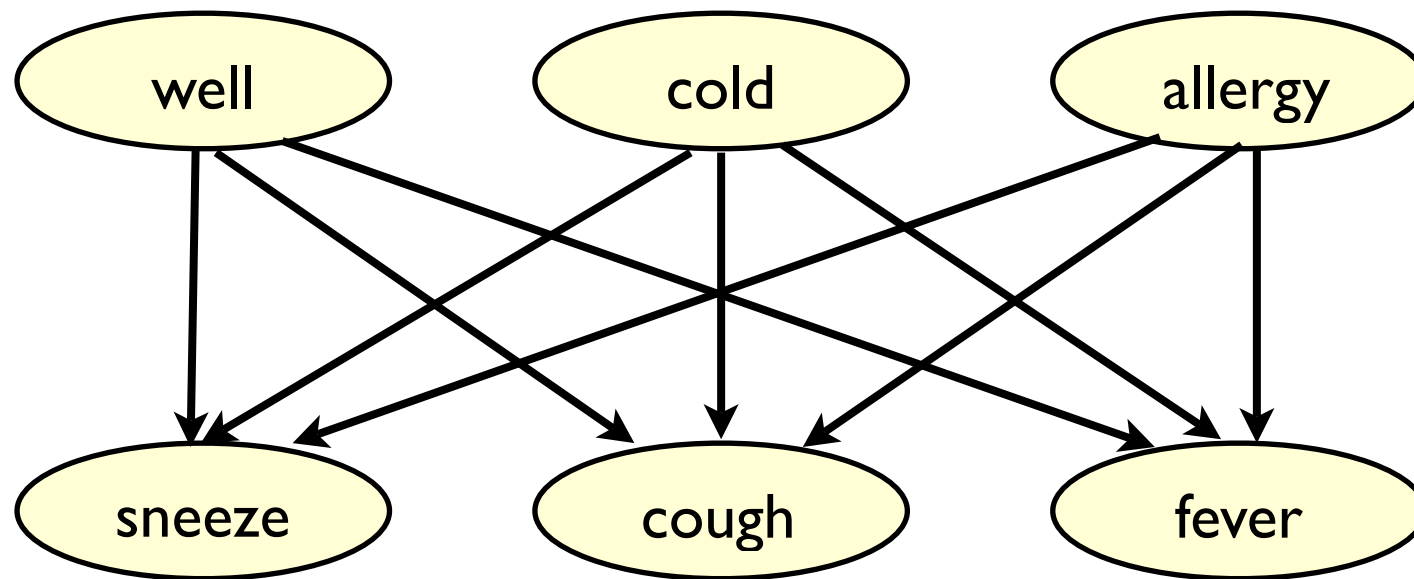
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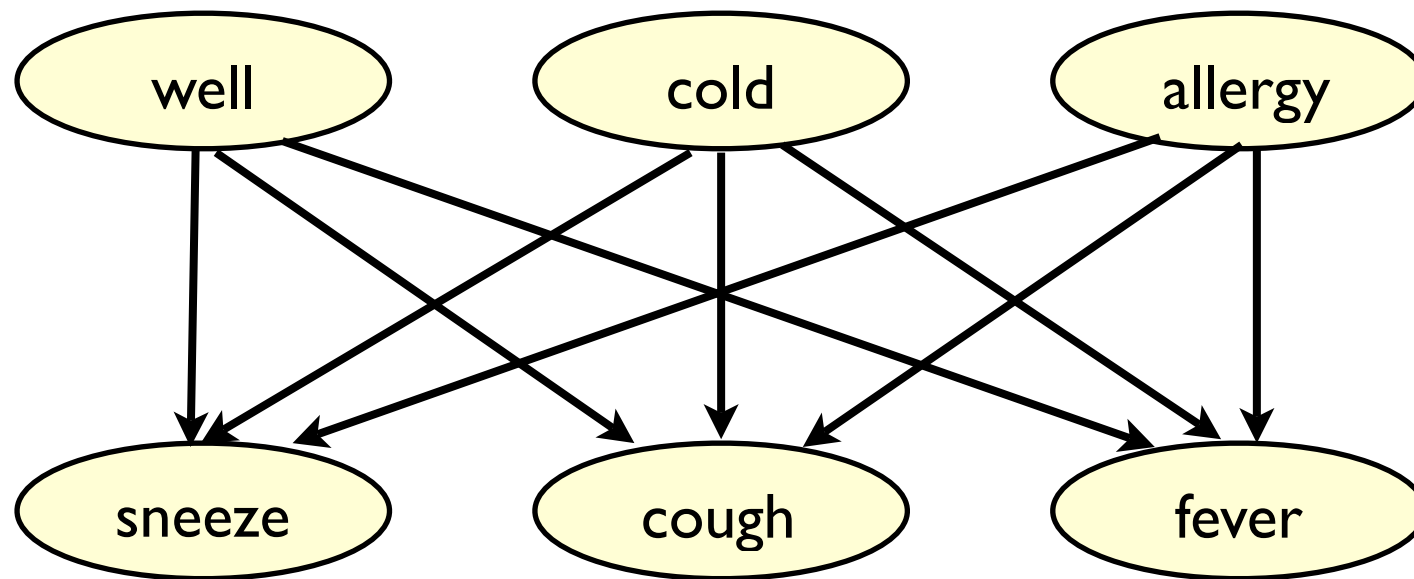
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Diagnosis	Well	Cold	Allergy
$P(C)$	0,90	0,05	0,05
$P(\text{sneeze} C)$	0,10	0,90	0,90
$P(\text{cough} C)$	0,10	0,80	0,70
$P(\text{fever} C)$	0,00	0,70	0,40

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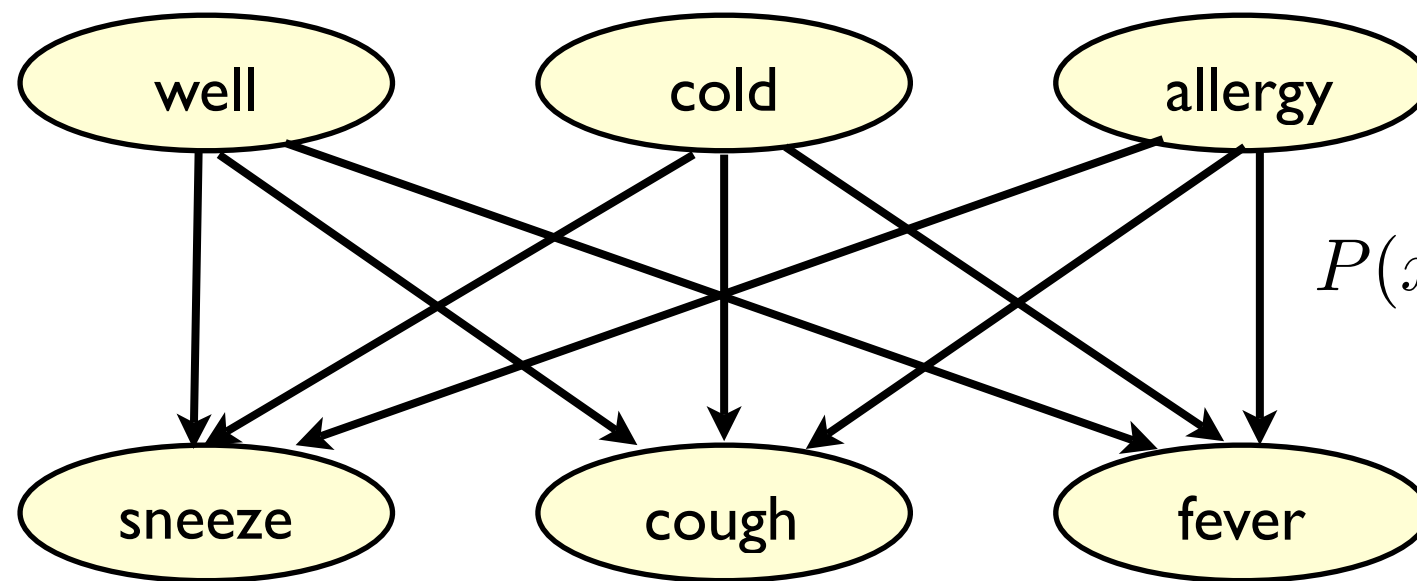


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Given that we observe  $x = \{\text{sneeze, cough, not fever}\}$   
 which class of diagnosis is most likely?

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

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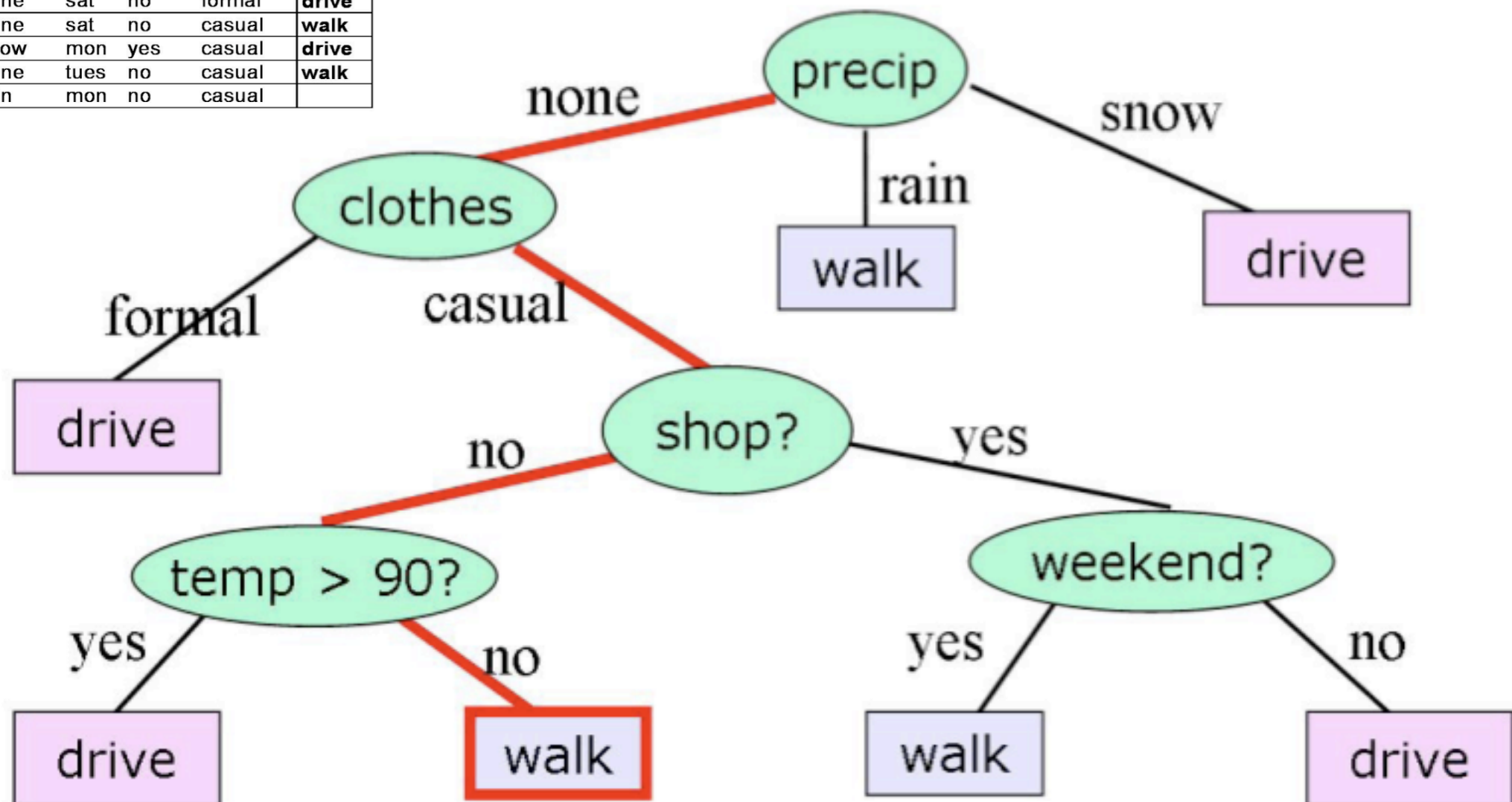
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<b>temp</b>	<b>precip</b>	<b>day</b>	<b>shop</b>	<b>clothes</b>	
71	none	fri	yes	formal	<b>drive</b>
36	none	sun	yes	casual	<b>walk</b>
62	rain	weds	no	casual	<b>walk</b>
93	none	mon	no	casual	<b>drive</b>
55	none	sat	no	formal	<b>drive</b>
80	none	sat	no	casual	<b>walk</b>
19	snow	mon	yes	casual	<b>drive</b>
65	none	tues	no	casual	<b>walk</b>
58	rain	mon	no	casual	

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# Prerequisites

- ✓ DM502, DM503 Programming (Programmering)
- ✓ DM527 Discrete Mathematics (Matematiske redskaber i datalogi)
- ✓ MM501 Calculus I
- ✓ DM509 Programming Languages (Programmeringssprog)
- ✓ ST501 Science Statistics (Science Statistik)

# Final Assessment (5 ECTS)

- ▶ A three hours written exam
  - closed book with a maximum of two two-sided sheets of notes.
  - external examiner
- ▶ 3 written and programming homeworks
  - pass/fail grading
  - internal examiner
  - [Prolog|Comet] (for 3.) and [Java|Python] and [R]

# Course Material

- ▶ Text book
  - Russell, S. & Norvig, P. *Artificial Intelligence: A Modern Approach*  
Prentice Hall, 2003
- ▶ Slides
- ▶ Source code and data sets
- ▶ [www.imada.sdu.dk/~marco/DM533](http://www.imada.sdu.dk/~marco/DM533)

