

# Knowledge Representation and Reasoning


*Artificial Intelligence Challenge / Introduction to Artificial Intelligence*

ICM 2A + M1 CPS<sup>2</sup>

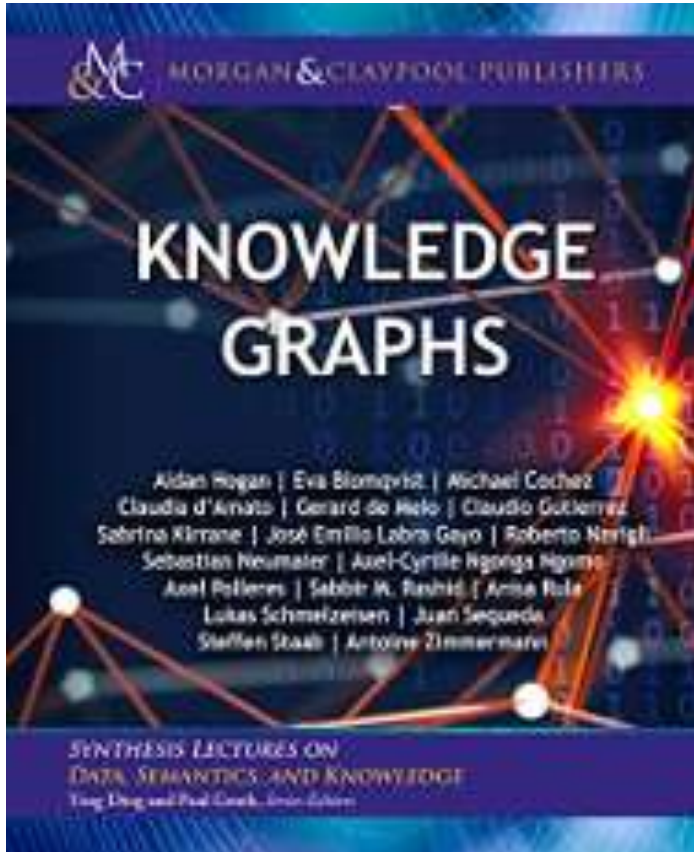
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[antoine.zimmermann@emse.fr](mailto:antoine.zimmermann@emse.fr)

# Organisation of the course

- One Web page where I put all the resources for this course:  
<https://www.emse.fr/~zimmermann/Teaching/KRR/>
- 6 sessions, including an exam, mixing *lectures*, *tutorials*, and *lab sessions*
- Focus on knowledge **representation** more than *reasoning techniques*
- Evaluation based on homework in groups and online MCQs
- I hope classwork can be interactive: student <-> student interactions as well as student <-> teacher interaction
- Bring your computers. You are **encouraged** to look up things on the Web when I'm presenting, but **no games**, **no social networks**, **no chat** during classes  and **keep your attention on what I'm saying**

# Book I'll use for this course



- Hogan et al. (2021): *Knowledge Graphs*. Morgan & Claypool.
- <https://kgbook.org/>

# What is *reasoning*?

- There are different forms of reasoning
  - Inductive reasoning
    - Build general theories (knowledge) from particular observations (facts)  
*e.g. If it has wheels, doors, seats, windows, engine, it must be a car.*
  - Deductive reasoning
    - Use general theories (knowledge) on particular cases to infer new facts  
*e.g. All cars are vehicles. Emmanuel Macron's DS 7 Crossback is a car. Therefore, Macron's DS 7 Crossback is a vehicle.*
  - Abductive reasoning
    - Use general theories and incomplete observations to get most plausible conclusion  
*e.g. My car is not in my garage. Most likely, my wife must still be at her work place.*  
*e.g. My car is not in my garage and my wife is at home. It may have been stolen!*

# Interactions between the types of reasoning

- Inductive reasoning can be used to devise the knowledge used later in deductive reasoning
- Deductive reasoning can be used to avoid inducing inconsistent knowledge or to validate inductive hypothesis (e.g., the Chinese hypothesis)
- Deductive reasoning can be used to check if an abductive reasoning is plausible
- Abduction can be used in combination with inductive reasoning to only induce the more plausible facts
- Etc.

# Inductive reasoning vs. deductive reasoning (1)

- Playing Go requires knowledge of the rules of the game
- From knowledge of the rules, one can decide if a stone placement is allowed or disallowed (*deductive reasoning*)
- One *could* learn Go from mere observation of games (*inductive reasoning*), but they would have a chance of making illegal placement, or may not attempt placements that are in fact legal
- It is **relatively easy** to formalise the rules of Go (with few corner cases that are trickier)
- It is **very difficult** to ensure that we know all legal/illegal placements by observation. It is even more difficult to define scoring this way

# Inductive reasoning vs. deductive reasoning (2)

- Winning a game of Go requires a lot of experience
- We do not know how to define general knowledge of the game such that the next move logically follows from our knowledge
- Playing good Go is better achieved by inductive reasoning from observation of a lot of games
- **But** players still need to know exactly the rules of the game to avoid making an illegal move

# Inductive reasoning and the Chinese hypothesis (1)

For a natural number  $n > 1$ , is the number  $2^n - 2$  divisible by  $n$  ?

$$2^2 - 2 = 2 = 1 \times 2$$

$$2^3 - 2 = 6 = 2 \times 3$$

$$2^4 - 2 = 14 = 2 \times 7$$

$$2^5 - 2 = 30 = 6 \times 5$$

$$2^6 - 2 = 62 = 2 \times 31$$

$$2^7 - 2 = 126 = 18 \times 7$$

$$2^8 - 2 = 254 = 2 \times 127$$

$$2^9 - 2 = 510 = 2 \times 3 \times 5 \times 17$$

$$2^{10} - 2 = 1\,022 = 2 \times 7 \times 73$$

$$2^{11} - 2 = 2\,046 = 186 \times 11$$

$$2^{12} - 2 = 4\,094 = 2 \times 23 \times 89$$

$$2^{13} - 2 = 8\,190 = 630 \times 13$$

$$2^{14} - 2 = 16\,382 = 2 \times 8\,191$$

$$2^{15} - 2 = 32\,766 = 2 \times 3 \times 43 \times 127$$

$$2^{16} - 2 = 65\,534 = 2 \times 7 \times 31 \times 151$$

$$2^{17} - 2 = 131\,070 = 7\,710 \times 17$$

$$2^{18} - 2 = 262\,142 = 2 \times 131\,071$$

$$2^{19} - 2 = 524\,286 = 27\,594 \times 19$$

$$2^{20} - 2 = 1\,048\,574 = 2 \times 524\,287$$



# Inductive reasoning and the Chinese hypothesis (2)

For a natural number  $n > 1$ , is the number  $2^n - 2$  divisible by  $n$  ?

21 X

22 X

23 ✓

24 X

26 X

27 X

28 X

29 ✓

30 X

31 ✓

32 X

33 X

34 X

35 X

36 X

37 ✓

38 X

39 X

40 X

41 ✓

42 X

43 ✓

44 X

45 X

46 X

47 ✓

48 X

49 X

# Inductive reasoning and the Chinese hypothesis (3)

- Is it the case that:

**for all prime number  $n$ , the number  $2^n - 2$  is divisible by  $n$**

???

# Inductive reasoning and the Chinese hypothesis (3)

- Is it the case that:

**for all prime number  $n$ , the number  $2^n - 2$  is divisible by  $n$**

???

Yes it is!

*Beware inductive reasoning*

*Machine learning is not good for everything*

# Inductive reasoning and the Chinese hypothesis (4)

- But, wait:

**Is the number  $2^{341} - 2$  divisible by 341**

**???**

# Inductive reasoning and the Chinese hypothesis (4)

- But, wait:

**Is the number  $2^{341} - 2$  divisible by 341**

**???**

**Yes it is!**

...but  $341 = 11 \times 31$  is not a prime

# Inductive reasoning vs. deductive reasoning (3)

- Deductive reasoning is crucial in situation when the **risk of failure** of inductive reasoning overcome the **benefit of probable success**
- Inductive reasoning is particularly good for tasks recognising a situation based on perception

*e.g., based on the intensity of light signals and my experience of seeing urban areas, I can say that the traffic light on the car lane of my car is red*

- Deductive reasoning is particularly good for tasks that require strict application of rules

*e.g., considering that I drive a car on a car lane where the traffic light is red, and I do not have a priority vehicle (ambulance, etc), I must stop*

# What is this course about?

- Traditionally, KR&R (and this course) is about the techniques to computerise **deductive reasoning**. In particular:
  - How can knowledge can be represented in machines to allow the computation of deduction? → *symbolic knowledge representation*
  - How to interpret the digital representation in terms of what it *means* → *formal semantics*
  - How to manipulate the digital representation of knowledge to automate reasoning? → *data structures and algorithms*
  - How to use these techniques adequately in concrete use cases? → *knowledge engineering*

# Requirements

- The foundations of KR&R are strongly related to logics
- Connections with EMSE 1st year course on *Introduction to logic*
  - The booklet of the logic course is available online (in French) at <https://www.emse.fr/~zimmermann/Teaching/Logique/Livret/>
- A good understanding of the basis of propositional and first order logics is *preferable*
- Basic programming information system skills are useful.



# Where is KR&R used?

These products embed, among other things, a knowledge base, and can do more or less advanced reasoning:

- virtual assistants (Alexa, Cortana, Siri, etc.)
- Wolfram Alpha
- Main search engines (the “[Google Knowledge Graph](#)” + [schema.org](#))  
→ consequently, web sites that want to be better indexed by search engines
- Enterprise Knowledge Graphs ([Airbnb](#), [Amazon](#), [eBay](#), [IBM](#), [LinkedIn](#), [Microsoft](#), [Uber](#), ...)
- Library data management systems (LoC, BNF, BNL, DNB, ...)
- [Wikidata](#)

# Highlight on *Knowledge Graphs*

- Knowledge Graph is a hot topic in AI that lies at the convergence of:
  - Knowledge Representation and Reasoning
  - Data management and processing (and Big Data)
  - Data analytics
  - Machine Learning (and graph embeddings)
  - The World Wide Web (and the Semantic Web)
  - (Linked) Open Data, FAIR scientific data (**F**indable, **A**ccessible, **I**nteroperable, **R**eusable data)

# MOOCS (in case Covid-19 keeps you at home)

- *CS 520: Knowledge Graphs.* (2021 course)  
<https://web.stanford.edu/~vinayc/kg/2021/>
  - *AI: Knowledge Representation and Reasoning.*  
<https://www.classcentral.com/course/swayam-ai-knowledge-representation-and-reasoning-7958>
  - *Knowledge Engineering with Semantic Web Technologies.*  
<https://open.hpi.de/courses/semanticweb2015>
  - *Knowledge Graphs.* (2020 course)  
<https://open.hpi.de/courses/knowledgegraphs2020>
  - *CS 227: Knowledge Representation and Reasoning.*  
<https://web.stanford.edu/class/cs227/> (no videos, just written resources)
- + There are more links to online resources on my web space

# Practical work: eCommerce knowledge graph

- What an intelligent agent needs to **know** to buy things online?
- What is it that I need?
- Where can I find it?
- How do I know something is what I need?
- How do I know something is for sale and I can buy it?
- How can I buy it?
- When I buy it, how to I get it?