



人工智能：知识表示与处理

赵一铮

南京大学人工智能学院



0

课程基本信息

Organizational

This course is taught by:

- 赵一铮 (zhaoyz@nju.edu.cn)

with TAs:

- 邓一鸣 (mg21370004@smail.nju.edu.cn)
- 王森 (mg21370034@smail.nju.edu.cn)
- 肖棹月 (mushmnot@gmail.com)

Prerequisites: some familiarity with:

- mathematical logics (esp. first-order logic)
- programming with basic Java

Teaching mode:

- lectures (online at Tencent Meeting for Weeks 1 and 2,
onsite at Yi-B 101 for the rest semester, **provisional**)

Organizational

Assessment:

- Assignments (50%)
- Final exam (50%)

Assignments released via PS:

- Five assignments in total, with each contributing 10%
- Immediately after lecture 3n ($n = 1, 2, 3, 4, 5$)

Solution submitted via PS:

- Due 2 weeks after each assignment release (e.g., 1st assignment to be released on 2nd March and solutions due on 16th March)
- Late submission (capped at 60%, unless having mitigating circumstances)

Marks & feedback given via PS

Assignments

Small questions:

- To ensure that you grasp the basics taught in lectures

Short essays of 200 – 500 words:

- To make you think
- To practice academic writing

Modeling tasks:

- To get your hands dirty
- To appreciate numerous ways in which things can be done

Programming tasks:

- To hone your Java programming skills
- Instructed stepwise by TAs and myself

Final Exam

Two hours – offline

EXAM PAPER MUST NOT BE REMOVED FROM THE EXAM ROOM

NANJING UNIVERSITY
SCHOOL OF ARTIFICIAL INTELLIGENCE

KNOWLEDGE REPRESENTATION AND PROCESSING

Date: Sunday, 27th June 2021

Time: 10:30 - 12:30

This is an offline examination.
The examination contains NINE base questions and ONE bonus question.
Be sure to answer at least ALL base questions.

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This is a CLOSED book examination
The use of electronic devices is NOT permitted

Question 10. Bisimulation (BONUS question)

Interpretations of ALC can be represented as graphs, with edges labelled by roles and nodes labelled by sets of concept names. More precisely, in such a graph

- each node corresponds to an element in the domain of the interpretation and it is labelled with all the concept names to which this element belongs in the interpretation;
- an edge with label r between two nodes says that the corresponding two elements of the interpretation are related by the role r .

Definition 1 (Bisimulation) Let \mathcal{I}_1 and \mathcal{I}_2 be interpretations. The relation $\otimes \subseteq \Delta^{\mathcal{I}_1} \times \Delta^{\mathcal{I}_2}$ is a bisimulation between \mathcal{I}_1 and \mathcal{I}_2 if:

- $d_1 \otimes d_2$ implies $d_1 \in A^{\mathcal{I}_1}$ iff $d_2 \in A^{\mathcal{I}_2}$, for any $d_1 \in \Delta^{\mathcal{I}_1}$, $d_2 \in \Delta^{\mathcal{I}_2}$, and A any concept name;
- $d_1 \otimes d_2$ and $(d_1, d'_1) \in r^{\mathcal{I}_1}$ implies the existence of $d'_2 \in \Delta^{\mathcal{I}_2}$ such that $d'_1 \otimes d'_2$ and $(d_2, d'_2) \in r^{\mathcal{I}_2}$, for any $d_1, d'_1 \in \Delta^{\mathcal{I}_1}$, $d_2 \in \Delta^{\mathcal{I}_2}$, and r any role name;
- $d_1 \otimes d_2$ and $(d_2, d'_2) \in r^{\mathcal{I}_2}$ implies the existence of $d'_1 \in \Delta^{\mathcal{I}_1}$ such that $d'_1 \otimes d'_2$ and $(d_1, d'_1) \in r^{\mathcal{I}_1}$, for any $d_1 \in \Delta^{\mathcal{I}_1}$, $d_2, d'_2 \in \Delta^{\mathcal{I}_2}$, and r any role name;

Given $d_1 \in \Delta^{\mathcal{I}_1}$ and $d_2 \in \Delta^{\mathcal{I}_2}$, we define $(\mathcal{I}_1, d_1) \sim (\mathcal{I}_2, d_2)$ if there is a bisimulation \otimes between \mathcal{I}_1 and \mathcal{I}_2 such that $d_1 \otimes d_2$, and say that $d_1 \in \mathcal{I}_1$ is bisimilar to $d_2 \in \mathcal{I}_2$.

Intuitively, d_1 and d_2 are bisimilar if (i) they belong to the same concept name and (ii) for each role name r , they have bisimilar r -successors.¹ Following the above definition, answer the questions below:

- (1) Determine whether $a_1 \in \mathcal{I}_1$ is bisimilar to $b_1 \in \mathcal{I}_2$ as depicted in Figure 1. Justify your answers. (2 mark)
 - (2) Determine whether $a_1 \in \mathcal{I}_1$ is bisimilar to $b_1 \in \mathcal{I}_2$ as depicted in Figure 2. Justify your answers. (2 mark)
 - (3) Determine whether $a_2 \in \mathcal{I}_1$ is bisimilar to $b_2 \in \mathcal{I}_2$ as depicted in Figure 3. Justify your answers. (2 mark)
- *: a_i, b_j ($1 \leq i \leq 4, 1 \leq j \leq 5$) are elements, M (Male) and F (Female) are concept names, and c (hasChild) is a role name.

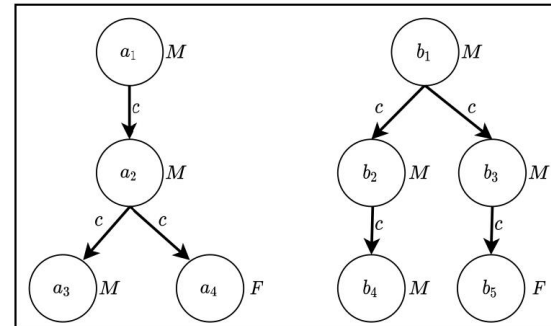


Figure 1: The interpretations \mathcal{I}_1 and \mathcal{I}_2 represented as graphs

¹An element q is called an r -successor of an element p if the two elements are related via the relation r such that $r(p, q)$



关于“定义”

Definitions

Extensional definitions:

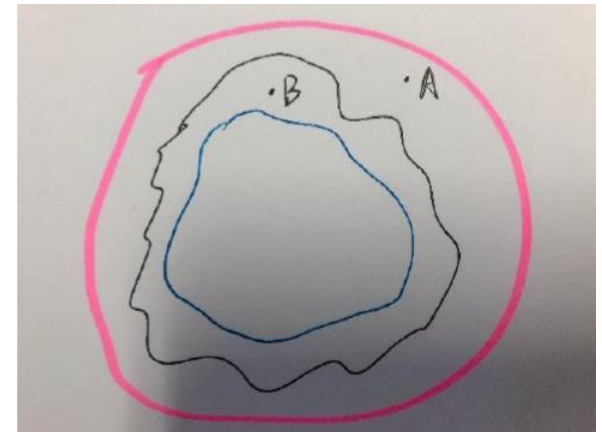
- Define a term by listing everything that falls under that definition
- E.g., **Permanent Five**: {China, France, Russia, UK, USA}

Intensional definitions:

- Define by specifying necessary and sufficient conditions for when the term should be used
- E.g., **Mother**: a female who has children

Mother: a female who has children

Mother: a female who has 2 children



1

基于“计算”的人工智能

What is artificial intelligence (AI)?

Intensional and imperfect definition of AI:

- the study of how to produce machines that have some of the qualities that **human mind** has, such as the ability to understand language, recognize pictures, solve problems, and learn
- Use artifacts (often machines) to simulate **intelligent behavior**

Definitions of intelligent behavior:

- Intensional: a group of general mental abilities
- Extensional: the ability to see, to hear, to smell, to feel, to think, to reason, to calculate, to communicate, to read, to understand, to imagine...to love
- Cannot find a perfect definition...

Definitions of intelligent behavior

Action (physical):

- the abilities to move, act...

Perception (physical & mental):

- the abilities to see, hear, smell, feel, or become aware of something through the senses (inputs)

Cognition (mental):

- the abilities to think, reason, calculate, communicate, or other mental processes of acquiring knowledge and understanding through thought, experience, and the senses (outputs)

Why AI is so important?

Generally speaking:

- to enhance the speed, precision and effectiveness of human efforts (to liberate productive forces and raise productivity)



Why AI is so powerful?

Generally speaking:

- Machines are **computationally** more powerful than humans, i.e., machines are more powerful for **computation** than humans
- Electrical energy superiority over bio energy
 - Faster processing of information (more efficient computation)
 - Better memory
 - Round-the-clock

Intensional definition of computation:

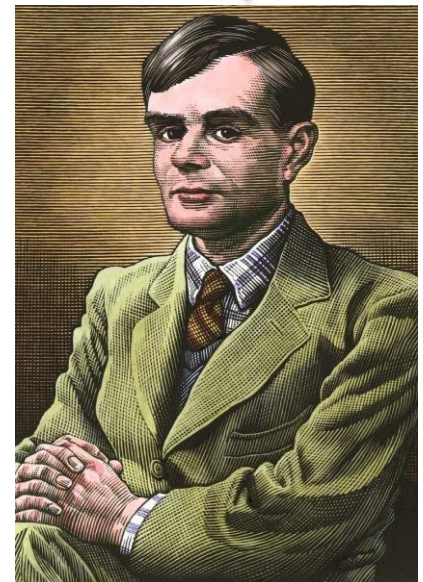
- Executes a set of instructions (called **program**) step by step
- Both arithmetical and non-arithmetical steps
- Relies heavily on mathematics

Subfields of AI

Yes, I posed questions but I proposed insights too!

To pass the **Turing Test**, a machine must possess (at least) the following capabilities:

1. to manipulate objects and move about (**robotics**)
2. to perceive objects (**computer vision**)
3. to allow it to communicate successfully in human language (**natural language processing**)
4. to store what it knows (**knowledge representation**)
5. to use the stored knowledge to answer questions and draw new conclusions (**automated reasoning**)
6. to adapt to new circumstances and to detect and extrapolate patterns (**machine learning**)



Pattern Recognition via Computation

An intelligent behavior:

- Input: an image



- Output: golden retriever (小金毛)
- Recognized in different ways:
 - via eyes and other bio-process for humans
 - via computation for machines

Language Understanding via Computation

An intelligent behavior:

- Input: a paragraph

The metal porch swing virtually sizzled on the old wooden front porch today. But we sat there anyway. Gramma wouldn't hear of anything else. I suggested a walk through the forest, hoping to entertain a breeze or two and to take advantage of the shade. Gramma shook her head. You were supposed to sit on the porch after supper, and that's what we were going to do.

- Question: This author implies that Gramma could not hear very well
- Output: no, but done in different ways:
 - via understanding and reasoning for humans
 - via computation for machines

All Done via Computation

Perception (physical & mental):

- the abilities to see, hear, smell, feel, or become aware of something through the senses (inputs – learning process)
- proper situations for machine learning

Cognition (mental):

- the abilities to think, reason, calculate, communicate, or other mental processes of acquiring knowledge and understanding through thought, experience, and the senses (outputs)

➤ Can machine learning do everything without help?

- hopefully yes, then we do not need to study this course, but unfortunately no☺



Limitations of Data-Driven AI

Consider a puzzle:

A Puzzle

You are on a strange island where people are divided into

- Knights – always saying the truth
- Knaves – always saying lies

You meet two natives of the island Alice and Bob, and ask them

“Are you knights or knaves?”

Alice answers *“At least one of us is a knave”*

What are Alice and Bob?

Limitations of Data-Driven AI

Consider a puzzle:

Alice: "At least one of us is a knave"

Alice



Bob



Alice



Bob



Alice



Bob



Alice



Bob



Limitations of Data-Driven AI

“I was lured into the world of machine learning while trying to discover the world of AI. I admit, it is exhilarating to make a computer do complex things to my liking without saying how it should do this. I ran riot in this world and have long forgotten my original goal whilst studying how to prepare data, engineer features and build deep networks”

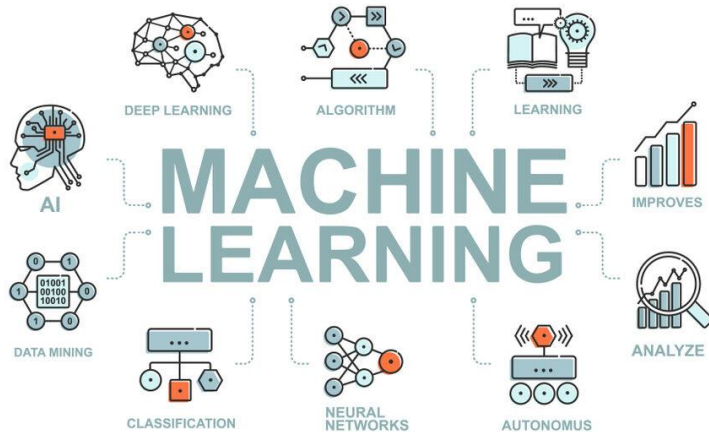
Main limitations:

- the issue of the **availability, quantity** and **quality** of associated data often remained wanting
- the increased acceptance of artificial intelligence in the industry gave rise to **fear, skepticism** and **resentment**
- People replaced by a **black box**, though the outputs seem to be identical (an **opaque box** is required, particularly in some areas)

2

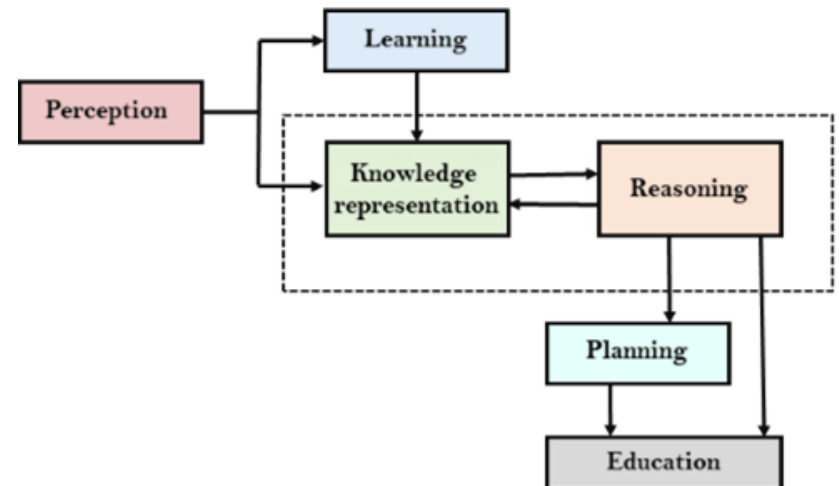
知识表示

Two Crucial Facets of Intelligence



The ability to **learn from experience** and perform better when confronted with similar situations or adapt to new situations

The ability to maintain an internal (abstract) state of knowledge and **reason over that knowledge** to draw new conclusions

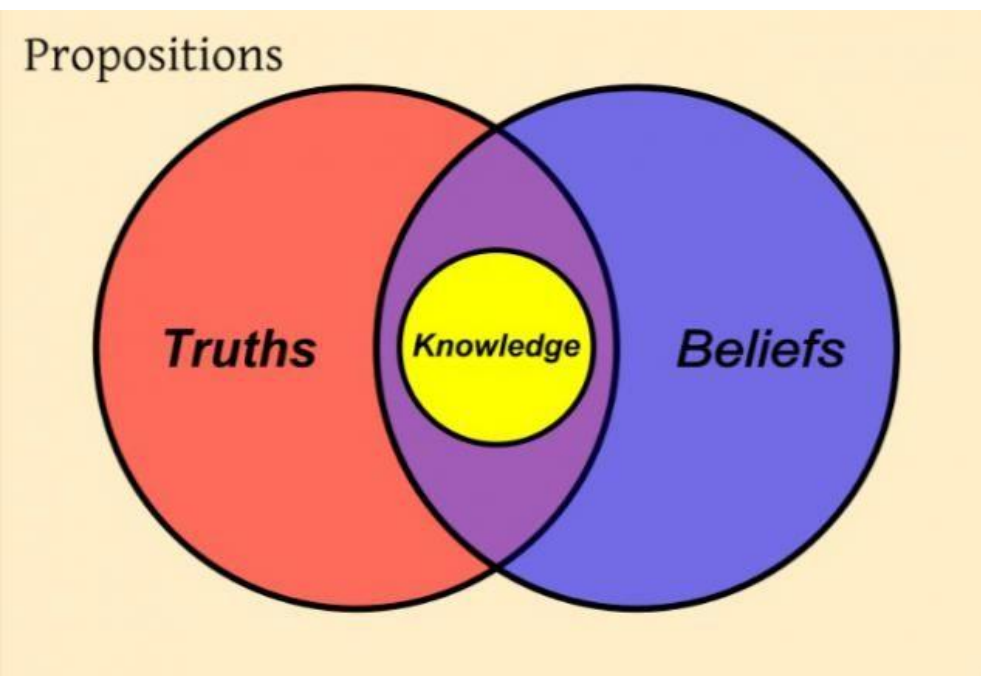


What is Knowledge?

“Knowledge”: dates back to Ancient Greek

- not totally demystified
- relation between knower and proposition

e.g. John knows that P = John knows that it is true that P



“knows” vs **“believes”**

- more prop. attitudes: **“hopes”**, **“regrets”**, **“doubts”**...

- **Idea**: John takes the world in one way, not another

What is Knowledge Representation?

“Representation”: as philosophically vexing as that of **“Knowledge”**

- not totally demystified
- relation between two domains
 - one domain (**representor**) stands for/takes place of another domain (**representee**)



- **representor**: more concrete, immediate, accessible in some way

How to Represent Knowledge?

LANGUAGE AND ITS ACQUISITION

“Natural Language” for
human beings

Therefore must be
“human processible”



Nigiri is a type of **Sushi** which has ingredient **Rice** and **Fish**

How to Represent Knowledge?

LANGUAGE AND ITS ACQUISITION

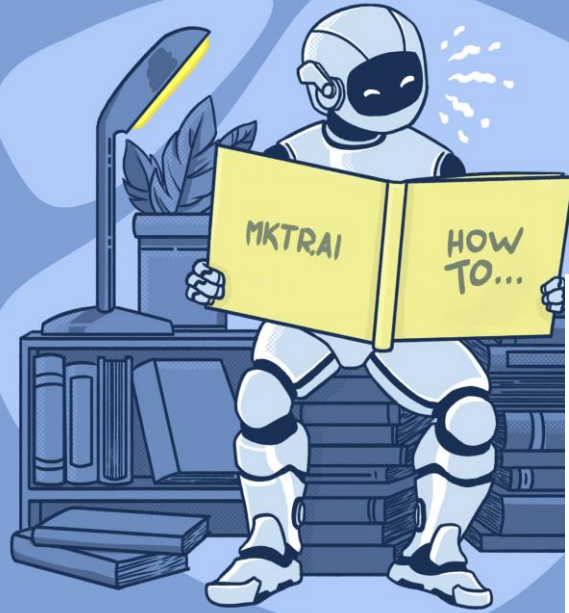
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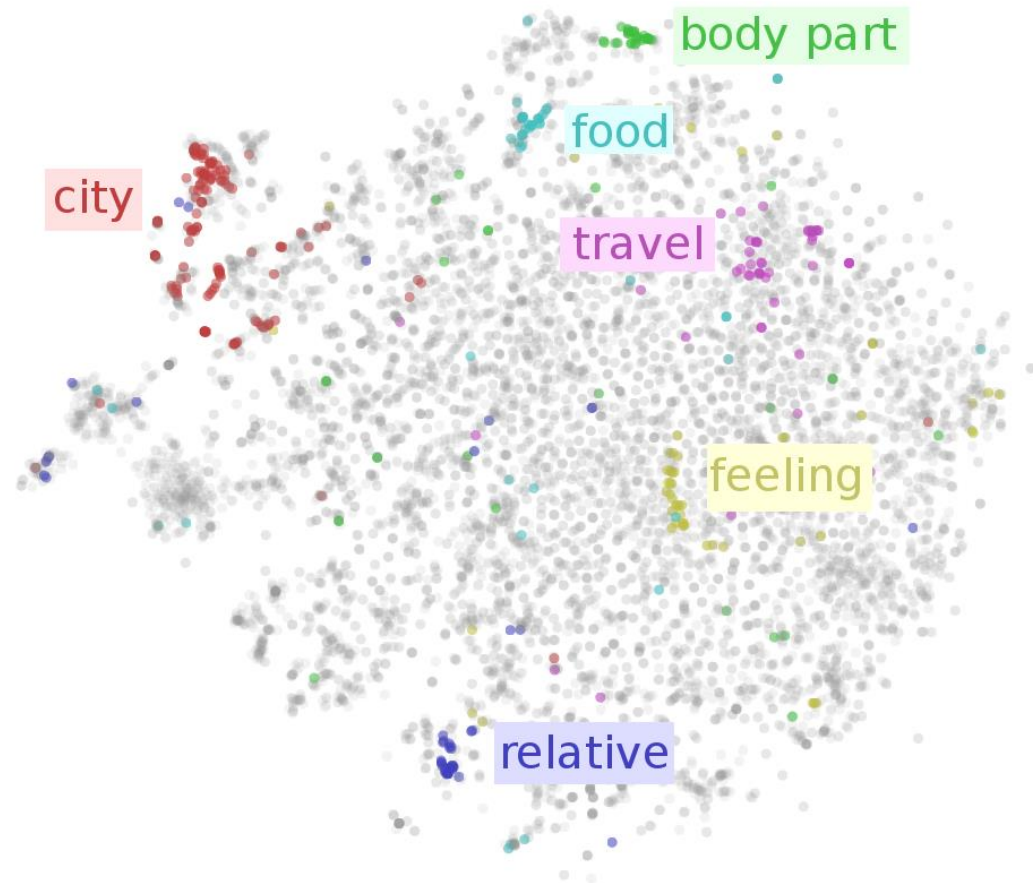
Nigiri is a type of **Sushi** which has ingredient **Rice** and **Fish**

How to Represent Knowledge?

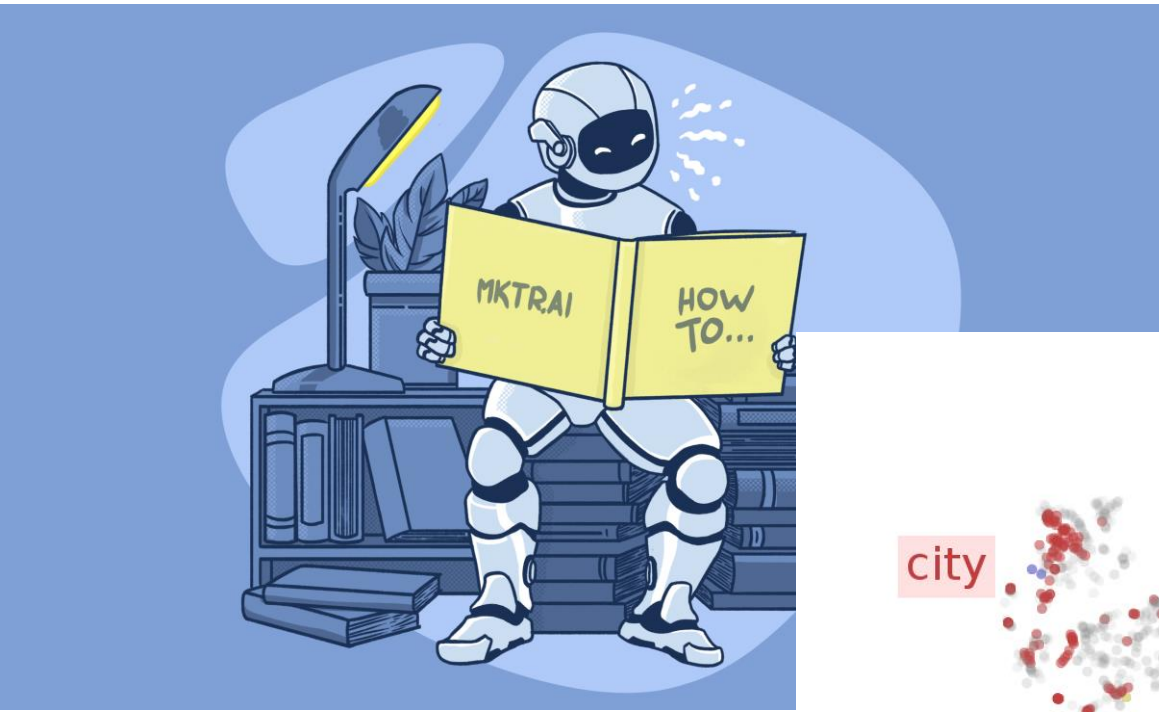


“computer processible”
representation

Computational Model



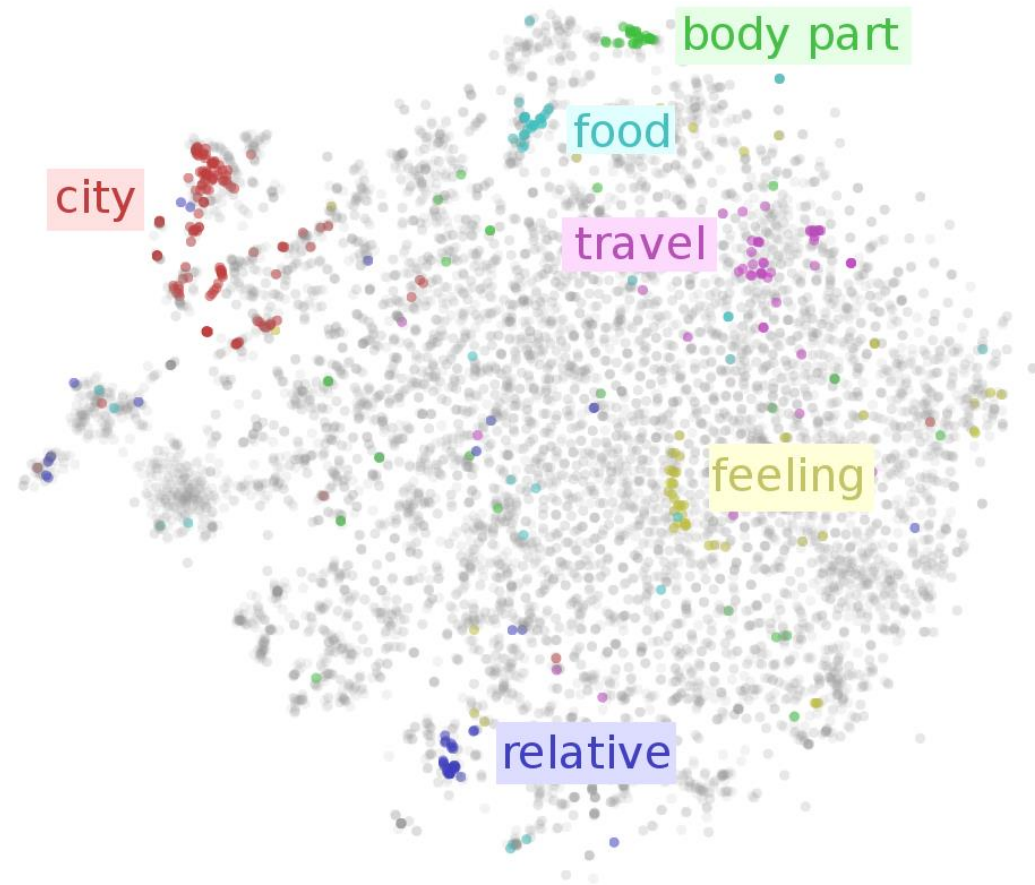
How to Represent Knowledge?



“computer processible”
representation

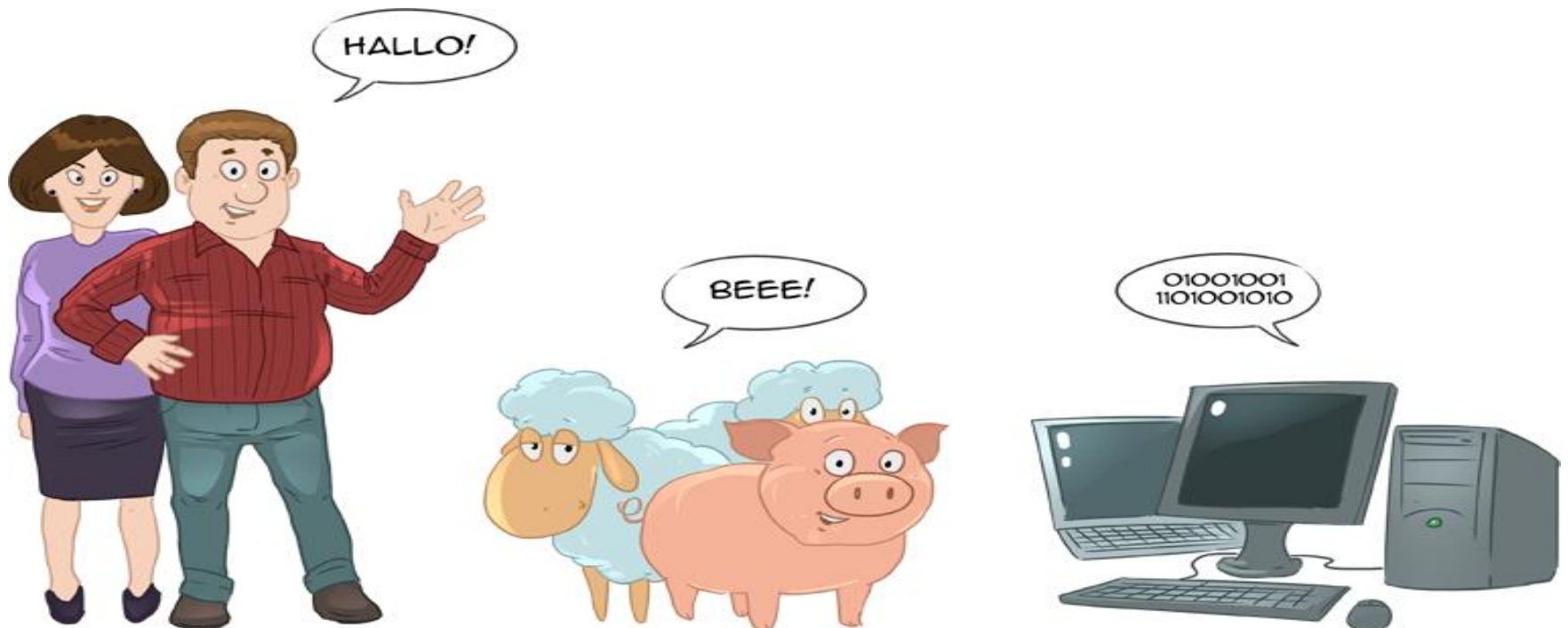
Computational Model

“word embedding”



How to Represent Knowledge?

Can we find **a KR language** that unifies the understanding of knowledge for both **human beings** and **computer beings**?

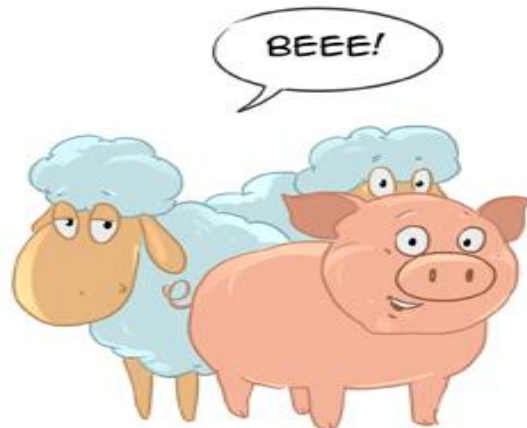


How to Represent Knowledge?

Can we find a
understanding
human beings



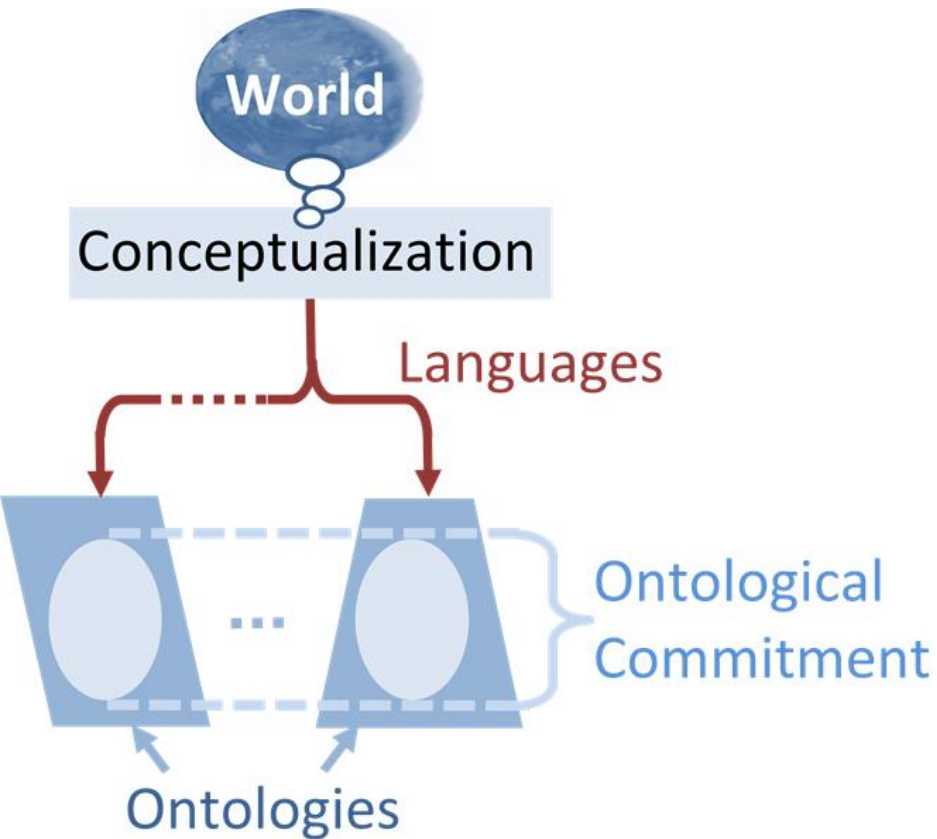
e that unifies the
ge for both
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3

本体与描述逻辑

What is an Ontology?

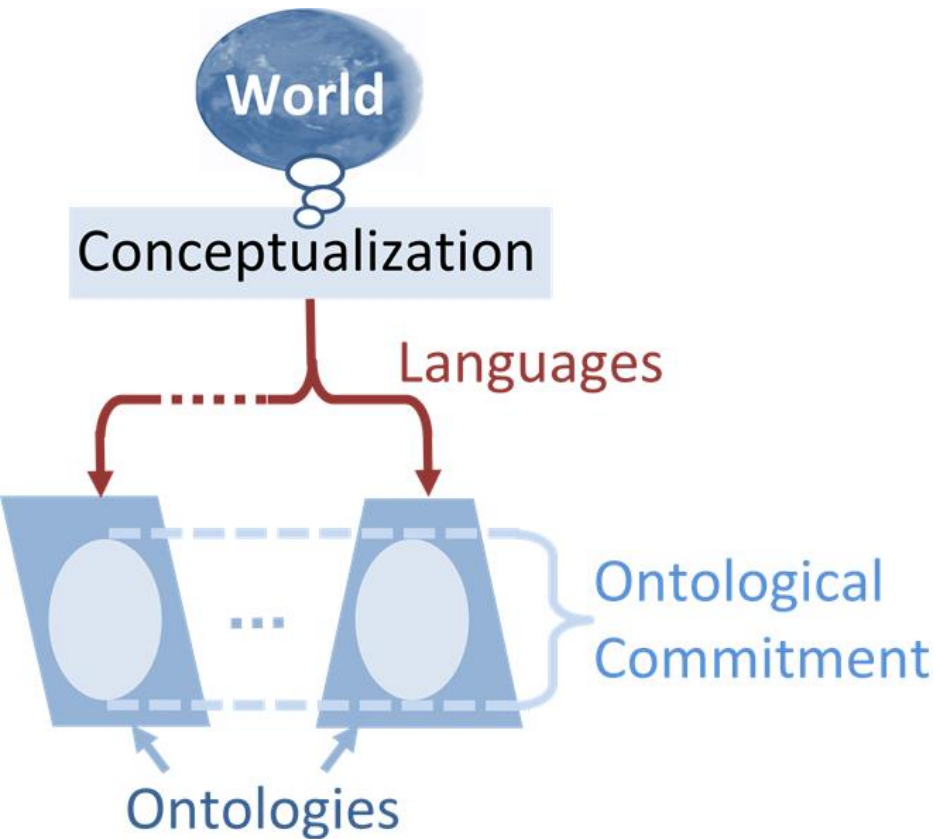


A **conceptualization** is an **abstract simplified view** of some selected part (domain) of the world

The **world**:
physical objects

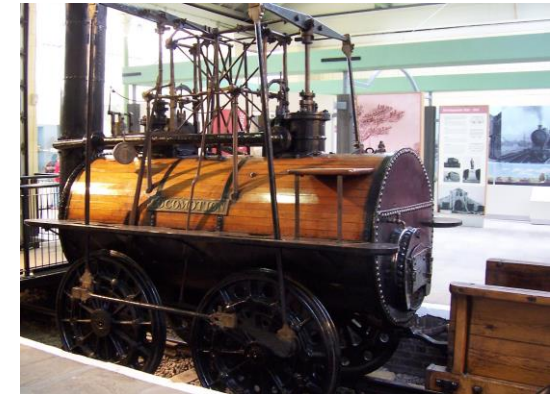
The **conceptualization**:
models of physical objects

What is an Ontology?

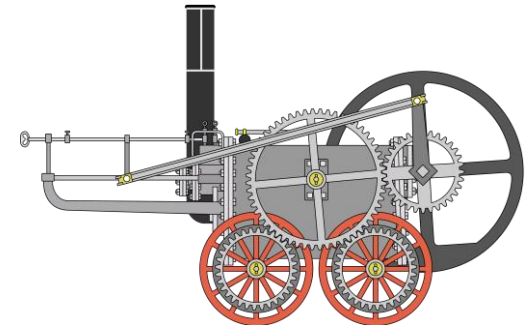


A **conceptualization** is an **abstract simplified view** of some selected part (domain) of the world

A **physical train** in real world

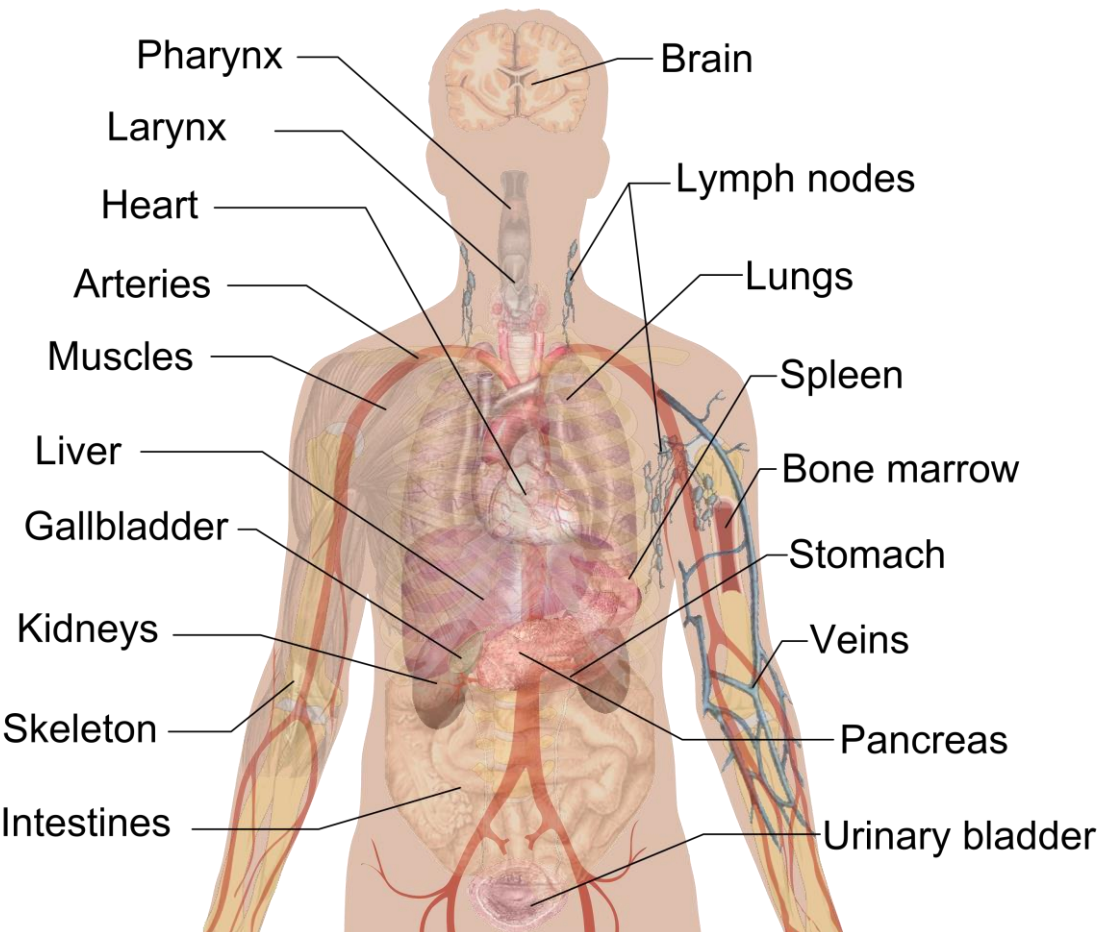


A **model** of the physical train in the conceptualization



What is an Ontology?

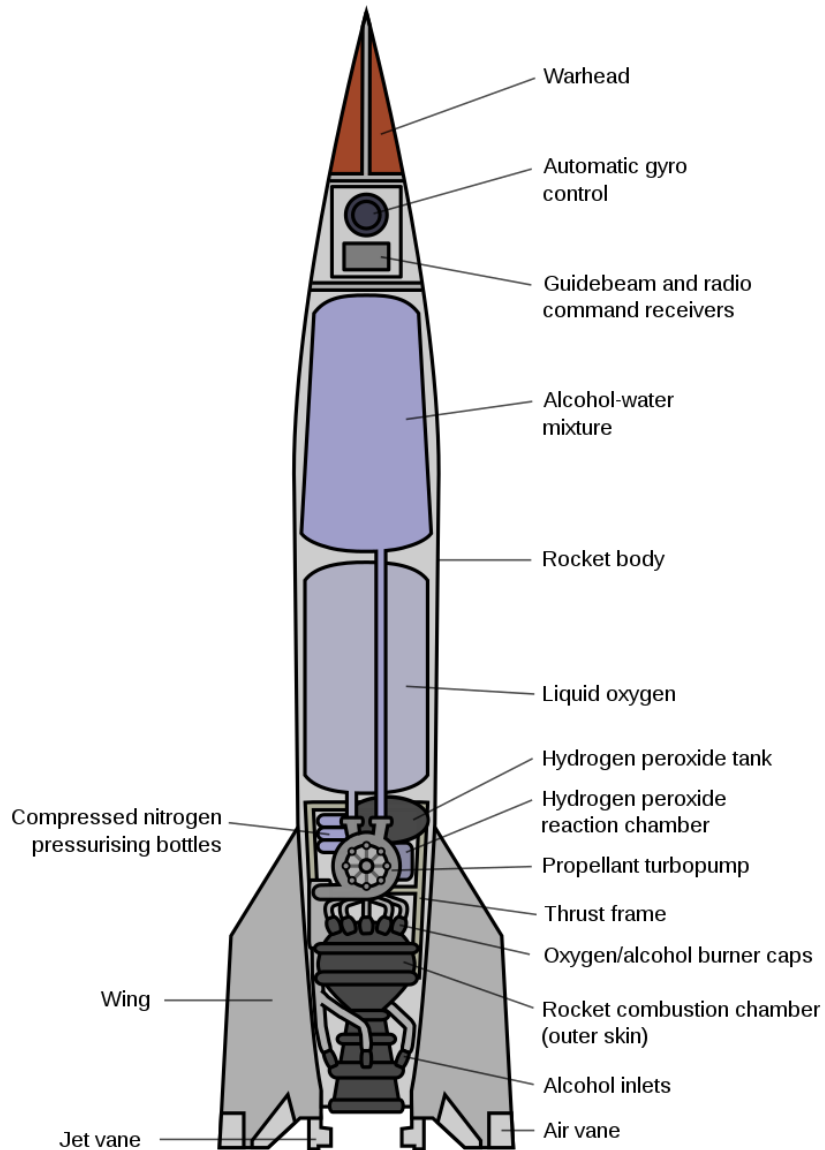
Internal organs



Fix vocabulary relevant
to domain, e.g.,

– **Anatomy**

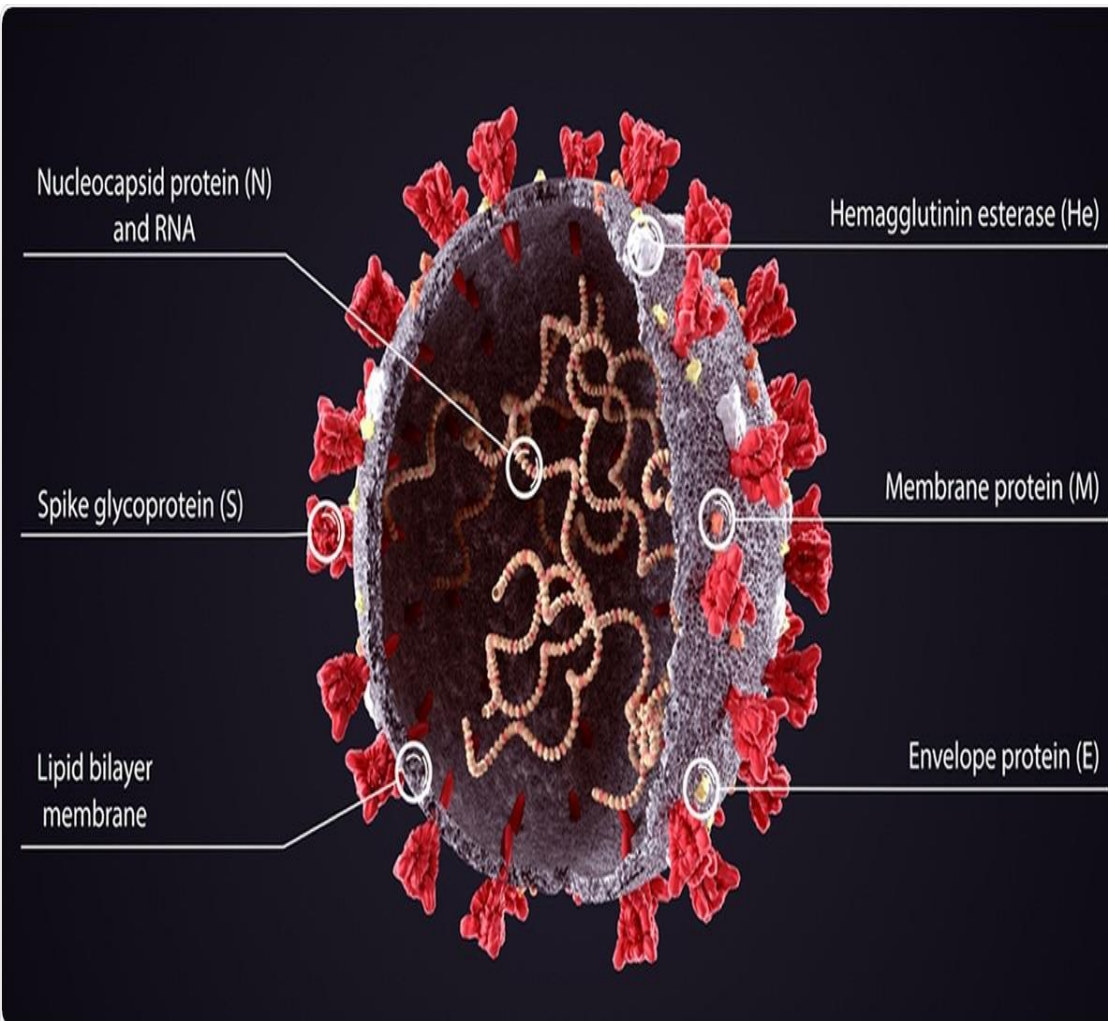
What is an Ontology?



Fix vocabulary relevant to domain, e.g.,

- Anatomy
- Rocket

What is an Ontology?



Fix vocabulary relevant to domain, e.g.,

- **Anatomy**
- **Rocket**
- **COVID-19**

What is an Ontology?

THE DIFFERENT TYPES OF SUSHI	
 <p>NIGIRI</p> <p>A topping (usually fresh fish) served on top of sushi rice</p>	 <p>URAMAKI</p> <p>Similar to a maki, but the rice is on the outside of the roll</p>
 <p>MAKI</p> <p>The sushi you probably see most often - rice and fillings wrapped in seaweed</p>	 <p>SASHIMI</p> <p>Fresh fish or shellfish served alone. Technically not sushi, but often served with it or mistaken for it</p>
 <p>INARI</p> <p>A fried tofu bean pouch filled with sushi rice. Can be plain or with a variety of toppings</p>	 <p>TEMAKI</p> <p>Sushi that has been hand-rolled into a cone shape</p>

Fix vocabulary relevant to domain

Fix meaning (semantics) of terms in vocabulary

Nigiri is a type of **Sushi** which has ingredient **Rice** and **Fish**

Ontology: A Computational Model?



Fix vocabulary relevant to domain

Fix meaning (semantics) of terms in vocabulary

Nigiri is a type of Sushi
which **has ingredient**
Rice and **Fish**

Specified in formal logic

$$\begin{aligned} \forall x. [\text{Nigiri}(x) \rightarrow \text{Sushi}(x) \wedge \\ \exists y. [\text{hasIngredient}(x, y) \wedge \\ \text{Rice}(y)] \wedge \\ \exists z. [\text{hasIngredient}(x, z) \wedge \\ \text{Fish}(z)]] \end{aligned}$$

Ontology: A Computational Model?



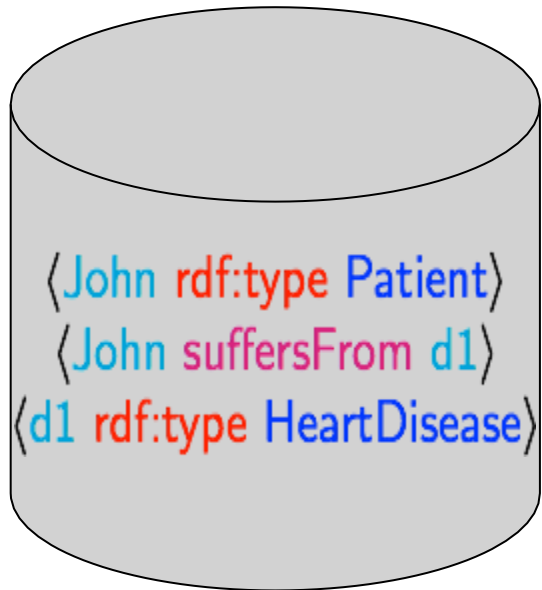
Fix vocabulary relevant to domain

Fix meaning (semantics) of terms in vocabulary

Nigiri is a type of Sushi which has ingredient Rice and Fish

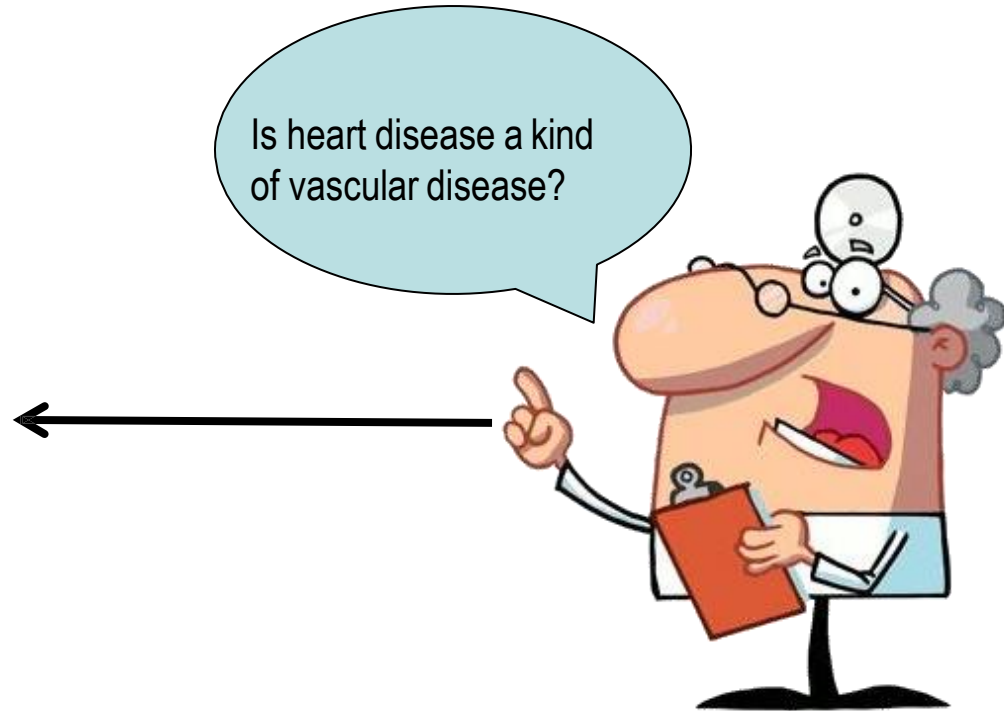
Natural language is NOT a computational model!

Standard Reasoning

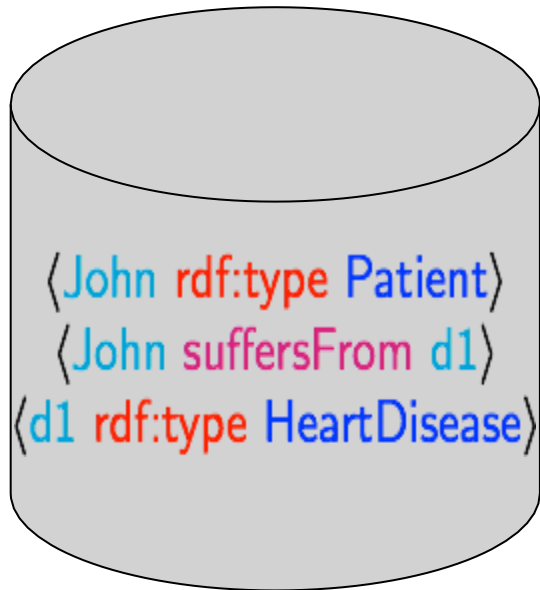


Heart \sqsubseteq MuscularOrgan \sqcap
 $\exists \text{isPartOf.CirculatorySystem}$
HeartDisease \equiv Disease \sqcap
 $\exists \text{affects.Heart}$
VascularDisease \equiv Disease \sqcap
 $\exists \text{affects.}(\exists \text{isPartOf.CirculatorySystem})$

Is heart disease a kind of vascular disease?



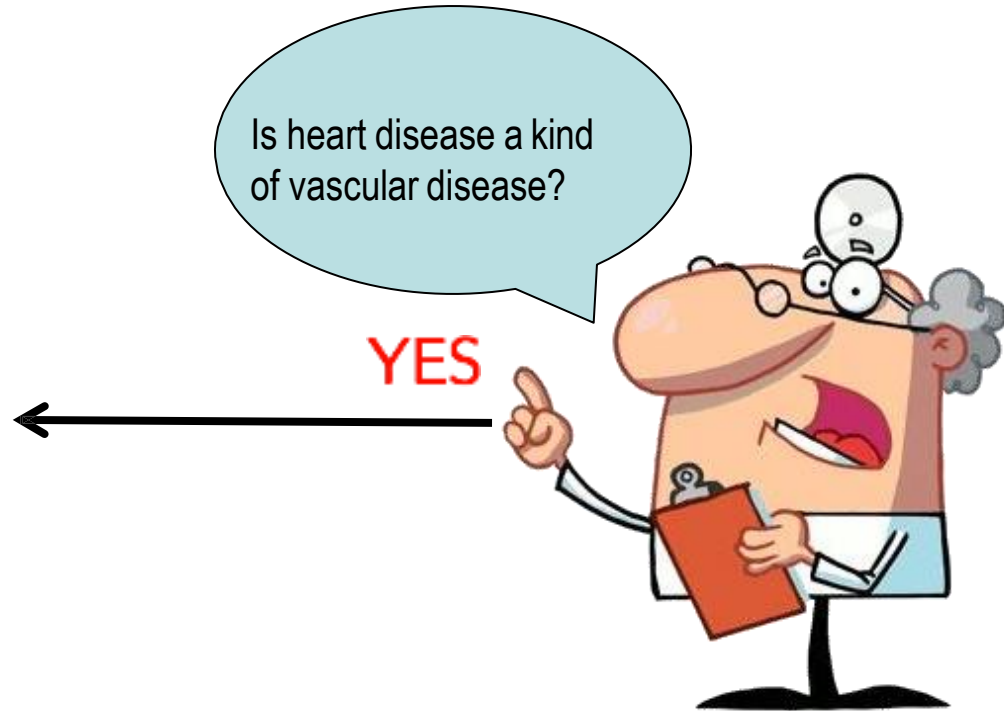
Standard Reasoning



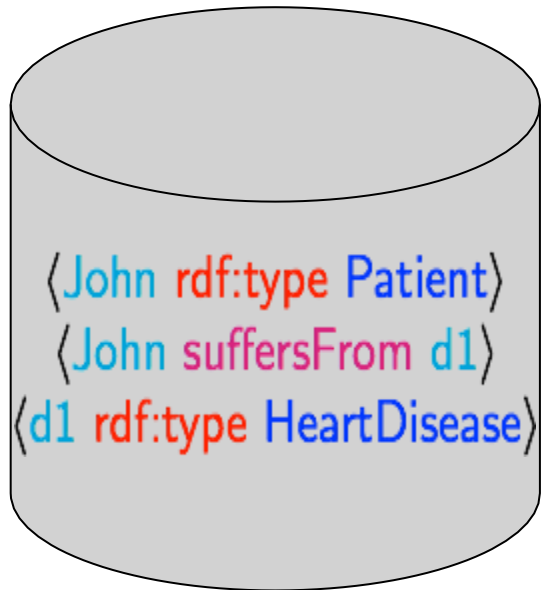
`Heart \sqsubseteq MuscularOrgan \sqcap`
 `\exists isPartOf.CirculatorySystem`
`HeartDisease \equiv Disease \sqcap`
 `\exists affects.Heart`
`VascularDisease \equiv Disease \sqcap`
 `\exists affects.(\exists isPartOf.CirculatorySystem)`

Is heart disease a kind of vascular disease?

YES



Non-Standard Reasoning



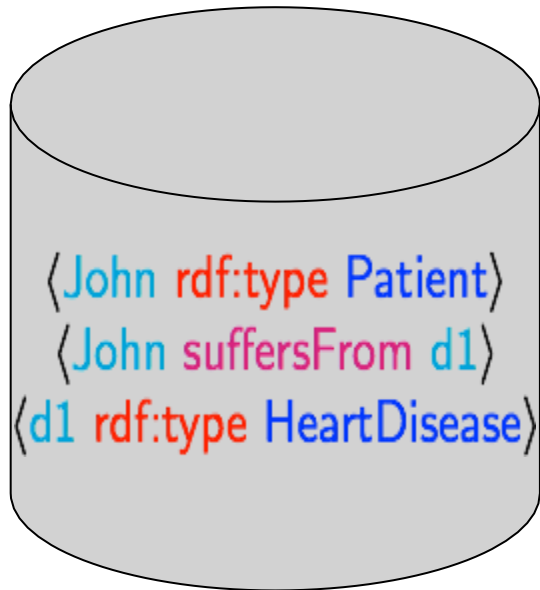
`<John rdf:type Patient>`
`<John suffersFrom d1>`
`<d1 rdf:type HeartDisease>`

`Heart \sqsubseteq MuscularOrgan \sqcap`
 `\exists isPartOf.CirculatorySystem`
`HeartDisease \equiv Disease \sqcap`
 `\exists affects.Heart`
`VascularDisease \equiv Disease \sqcap`
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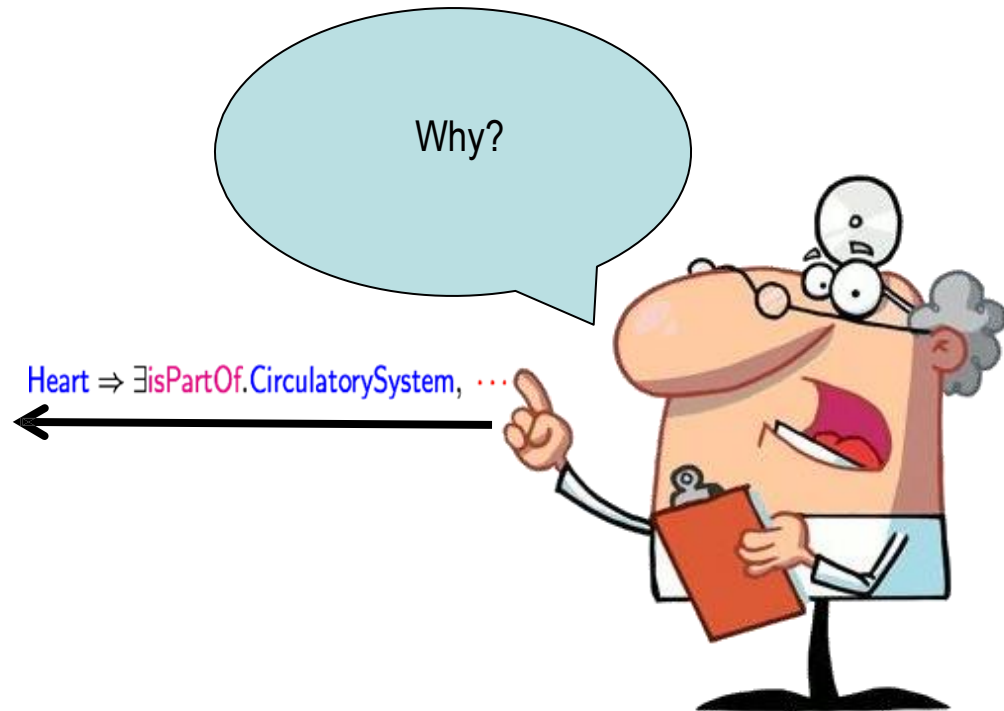
Why?



Non-Standard Reasoning



$\text{Heart} \sqsubseteq \text{MuscularOrgan} \sqcap \exists \text{isPartOf.CirculatorySystem}$
 $\text{HeartDisease} \equiv \text{Disease} \sqcap \exists \text{affects.Heart}$
 $\text{VascularDisease} \equiv \text{Disease} \sqcap \exists \text{affects} . (\exists \text{isPartOf.CirculatorySystem})$



Description Logics (DLs)

- Fragments of **first order logic** designed for KR
- Desirable computational properties
 - **Decidable** (essential)
 - Low complexity (desirable)
- Succinct and **variable free syntax**

Description Logics (DLs)

- Fragments of **first order logic (FOL)** designed for KR
- Desirable computational properties
 - **Decidable** (essential)
 - Low complexity (desirable)
- Succinct and **variable free syntax**

FOL:

$$\forall x. [\text{Nigiri}(x) \rightarrow \text{Sushi}(x) \wedge \exists y. [\text{hasIngredient}(x,y) \wedge \text{Rice}(y)] \wedge \exists z. [\text{hasIngredient}(x,z) \wedge \text{Fish}(z)]]$$

DL:

Nigiri \sqsubseteq **Sushi** \sqcap \exists **hasIngredient.Rice** \sqcap \exists **hasIngredient.Fish**

Description Logics (DLs)

DL Knowledge Base (KB) consists of two parts:

- Ontology (aka **TBox**) axioms define terminology (schema)

Nigiri \sqsubseteq **Sushi** \sqcap

\exists **hasIngredient.Rice** \sqcap

\exists **hasIngredient.Fish** \sqcap

- Ground facts (aka **ABox**) use the terminology (data)

messi : **Person** \sqcap

\exists **hasOccupation.Footballer**

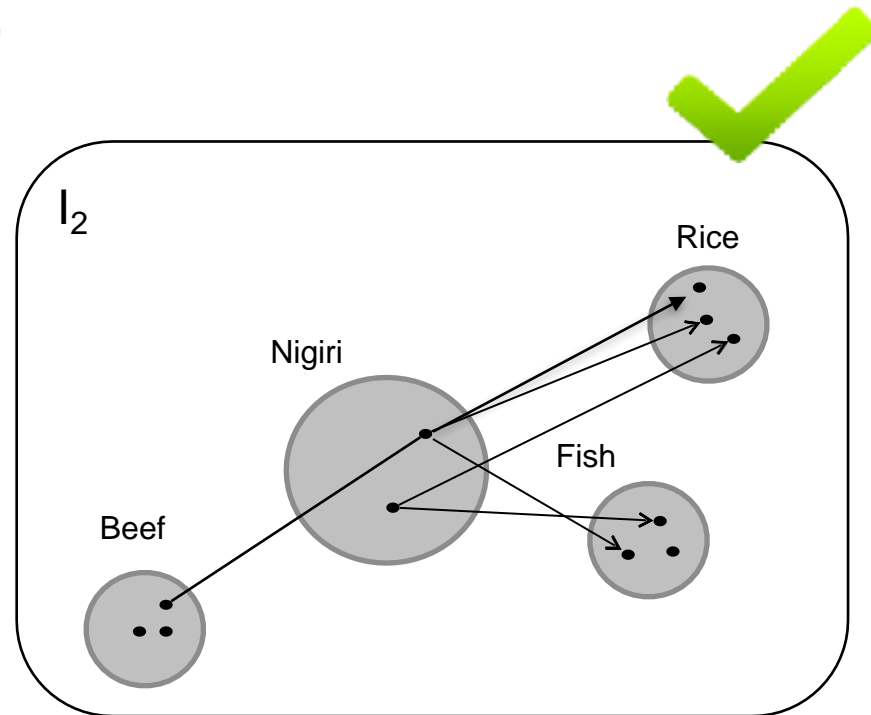
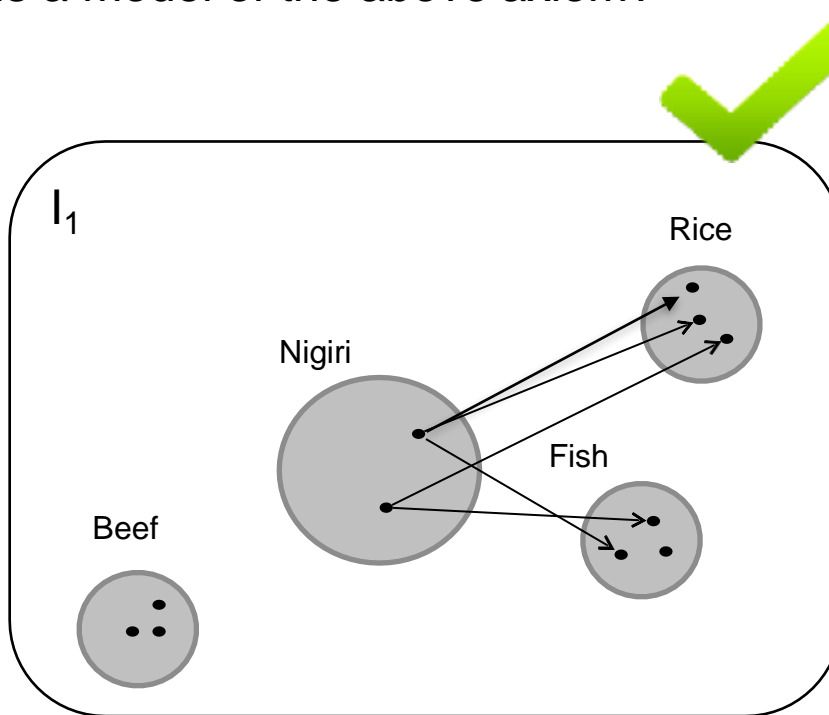
Model-Theoretic Semantics

Nigiri \sqsubseteq **Sushi** \sqcap

\exists **hasIngredient.Rice** \sqcap

\exists **hasIngredient.Fish**

Which of these interpretations
is a model of the above axiom?



\longrightarrow hasIngredient

Model-Theoretic Semantics

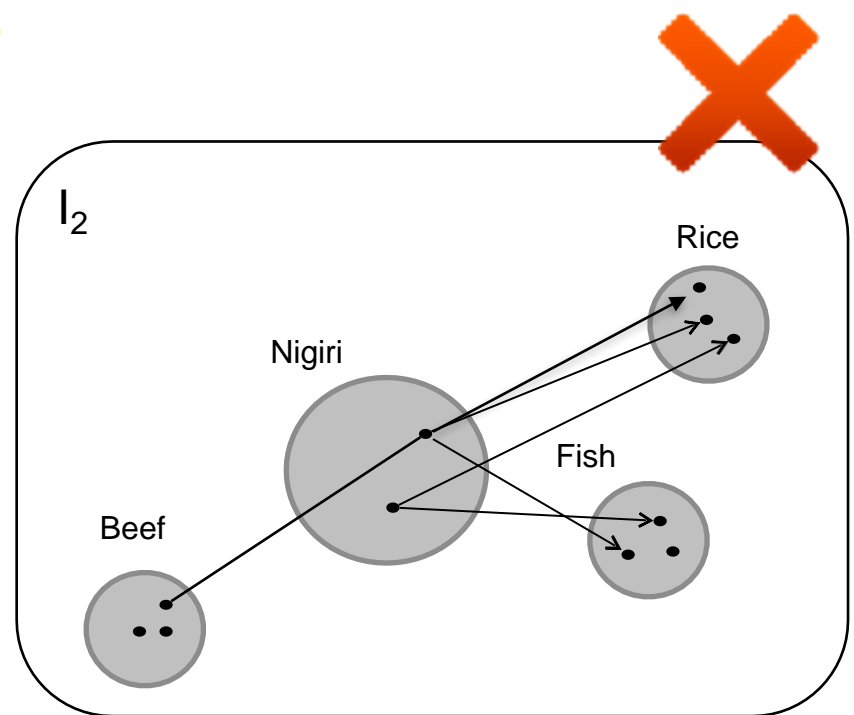
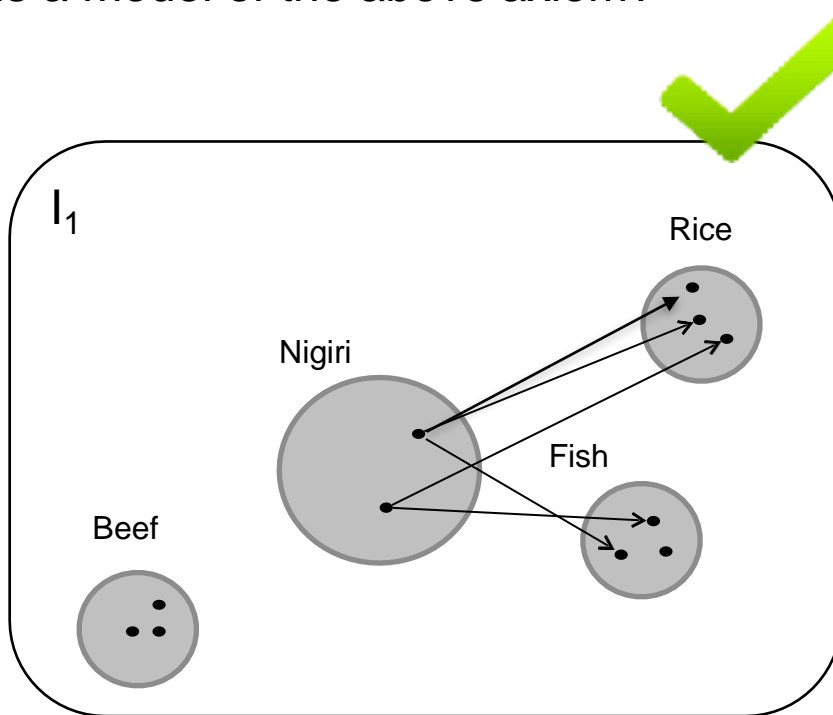
Nigiri \sqsubseteq **Sushi** \sqcap

\exists **hasIngredient**. **Rice** \sqcap

\exists **hasIngredient**. **Fish** \sqcap

\forall **hasIngredient**. (**Rice** \sqcup **Fish**)

Which of these interpretations
is a model of the above axiom?



\longrightarrow **hasIngredient**

3

基于逻辑语义的本体推理

Research Problems

To find proper logical languages for KR

- **Expressiveness**
more expressive power comes with computational cost
- **Computational properties**
 - decidability
e.g. first-order logic is undecidable
 - complexity
polynomial or exponential or even worse?



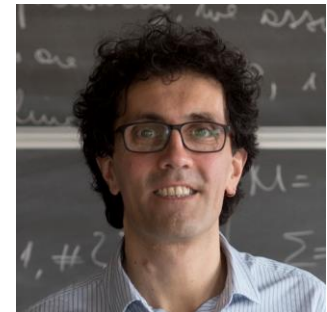
Franz Baader
TU Dresden



Carsten Lutz
Univ. of Bremen



Frank Wolter
Univ. of Liverpool



Diego Calvanese
Free Univ. of Bozen-Bolzano

Research Problems

To develop reasoning methods for KR languages

- **Standard reasoning (entailment checking)**

e.g. Boolean queries, check if $KB \models \alpha$

- **Non-standard reasoning**

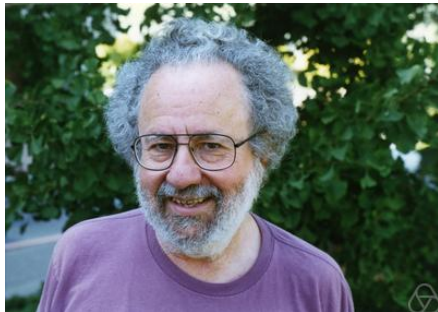
derive new knowledge from the existing, $KB \models KB'$

e.g. Jack has a child who is a lawyer

Jack has a child who is not a lawyer

What can you derive from these?

usually harder than standard reasoning



Martin Davis
New York Univ.



J. A. Robinson
Syracuse University



Ian Horrocks
Univ. of Oxford



Uli Sattler
The Univ. of Manchester

Standard Reasoning Tools

Major benefit of using DLs has been huge increase in range and sophistication of tools and infrastructure:

- Editors/development environments
- Reasoners

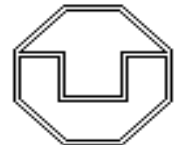
 **Hermit**

 **Racer**

 **KAON2**

 **FaCT++**

 **Pellet**

 **CEL**