

# **Rigorously Speaking, What are We?**

Bringsjord v. Granger

# Some Roots of the Debate

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 Theoretical Computer Science 

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## The grammar of mammalian brain capacity

A. Rodriguez, R. Granger\*

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**ABSTRACT**

Uniquely human abilities may arise from special-purpose brain circuitry, or from concerted general capacity increases due to our outsized brains. We forward a novel hypothesis of the relation between computational capacity and brain size, linking mathematical formalisms of grammars with the allometric increases in cortical-subcortical ratios that arise in large brains. In sum, i) thalamocortical loops compute formal grammars; ii) successive cortical regions describe grammar rewrite rules of increasing size; iii) cortical-subcortical ratios determine the quantity of stacks in single-stack pushdown grammars; iv) quantitative increase of stacks yields grammars with qualitatively increased computational power. We arrive at the specific conjecture that human brain capacity is equivalent to that of indexed grammars – far short of full Turing-computable (recursively enumerable) systems. The work provides a candidate explanatory account of a range of existing human and animal data, addressing longstanding questions of how repeated similar brain algorithms can be successfully applied to apparently dissimilar computational tasks (e.g., perceptual versus cognitive, phonological versus syntactic); and how quantitative increases to brains can confer qualitative changes to their computational repertoire.

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### 1. Brain growth shows surprisingly few signs of evolutionary pressure

Different animals exhibit different mental and behavioral abilities, but it is not known which abilities arise from specializations in the brain, i.e., circuitry to specifically support or enable particular capacities. Evolutionary constraints on brain construction severely narrow the search for candidate specializations. Although mammalian brain sizes span four orders of magnitude [1], the range of structural variation differentiating those brains is extraordinarily limited.

An animal's brain size can be roughly calculated from its body size [2], but much more telling is the relationship between the sizes of brains and of their constituent parts: the size of almost every component brain circuit can be computed with remarkable accuracy just from the overall size of that brain [1,3–5], and thus the ratios among brain parts (e.g. cortical to subcortical size ratios) increase in a strictly predictable allometric fashion as overall brain size increases [6,7] (Fig. 1).

These allometric regularities obtain even at the level of individual brain structures (e.g., hippocampus, basal ganglia, cortical areas). There are a few specific exceptions to the well-documented allometric rule (such as the primate olfactory system [8]), clearly demonstrating that at least some brain structure sizes can be differentially regulated in evolution, yet despite this capability, it is extremely rare for telencephalic structures ever to diverge from the allometric rule [4,6,7,9]. Area 10, the frontal pole, is the most disproportionately expanded structure in the human brain, and has sometimes been argued to be selected for differential expansion, yet the evidence has strongly indicated that area 10 (and the rest of anterior cortex) are nonetheless precisely the size that is predicted allometrically [6,7,10,11].

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## The modal argument for hypercomputing minds

Selmer Bringsjord\*, Konstantine Arkoudas

Department of Computer Science, Department of Cognitive Science, Rensselaer AI & Reasoning Laboratory, Rensselaer Polytechnic Institute (RPI), Troy, NY 12180, USA

Received 14 July 2003; received in revised form 21 October 2003

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**Abstract**

We now know both that hypercomputation (or super-recursive computation) is mathematically well-understood, and that it provides a theory that according to some accounts for some real-life computation (e.g., operating systems that, unlike Turing machines, never simply output an answer and halt) better than the standard theory of computation at and below the “Turing Limit.” But one of the things we do not know is whether the human mind hypercomputes, or merely computes—this despite informal arguments from Gödel, Lucas, Penrose and others for the view that, in light of incompleteness theorems, the human mind has powers exceeding those of TMs and their equivalents. All these arguments fail; their fatal flaws have been repeatedly exposed in the literature. However, we give herein a novel, formal *modal* argument showing that since it's mathematically possible that human minds are hypercomputers, such minds *are* in fact hypercomputers. We take considerable pains to anticipate and rebut objections to this argument. © 2003 Elsevier B.V. All rights reserved.

**Keywords:** Computationalism; Hypercomputation; Incompleteness theorems

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### 1. Introduction

Four decades ago, Lucas [50] expressed supreme confidence that Gödel's first incompleteness theorem (=Gödel I) entails the falsity of computationalism, the view that human persons are computing machines (e.g., Turing machines). Put barbarically, Lucas' basic idea is that minds are more powerful than Turing machines. Today, given our understanding of hypercomputation in theoretical computer science, and given the absolute consensus reigning in cognitive science that the human mind is, at least in large part, *some* sort of information-processing device, we know enough to infer that if Lucas is right, the mind is a hypercomputer. However, Lucas' arguments have

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The grammar of mammalian brain capacity 

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**Granger:**  
**We're less than a Turing machine!**

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The modal argument for hypercomputing minds

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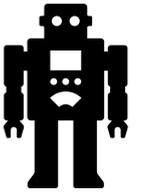
**Keywords:** Computationalism; Hypercomputation; Incompleteness theorems

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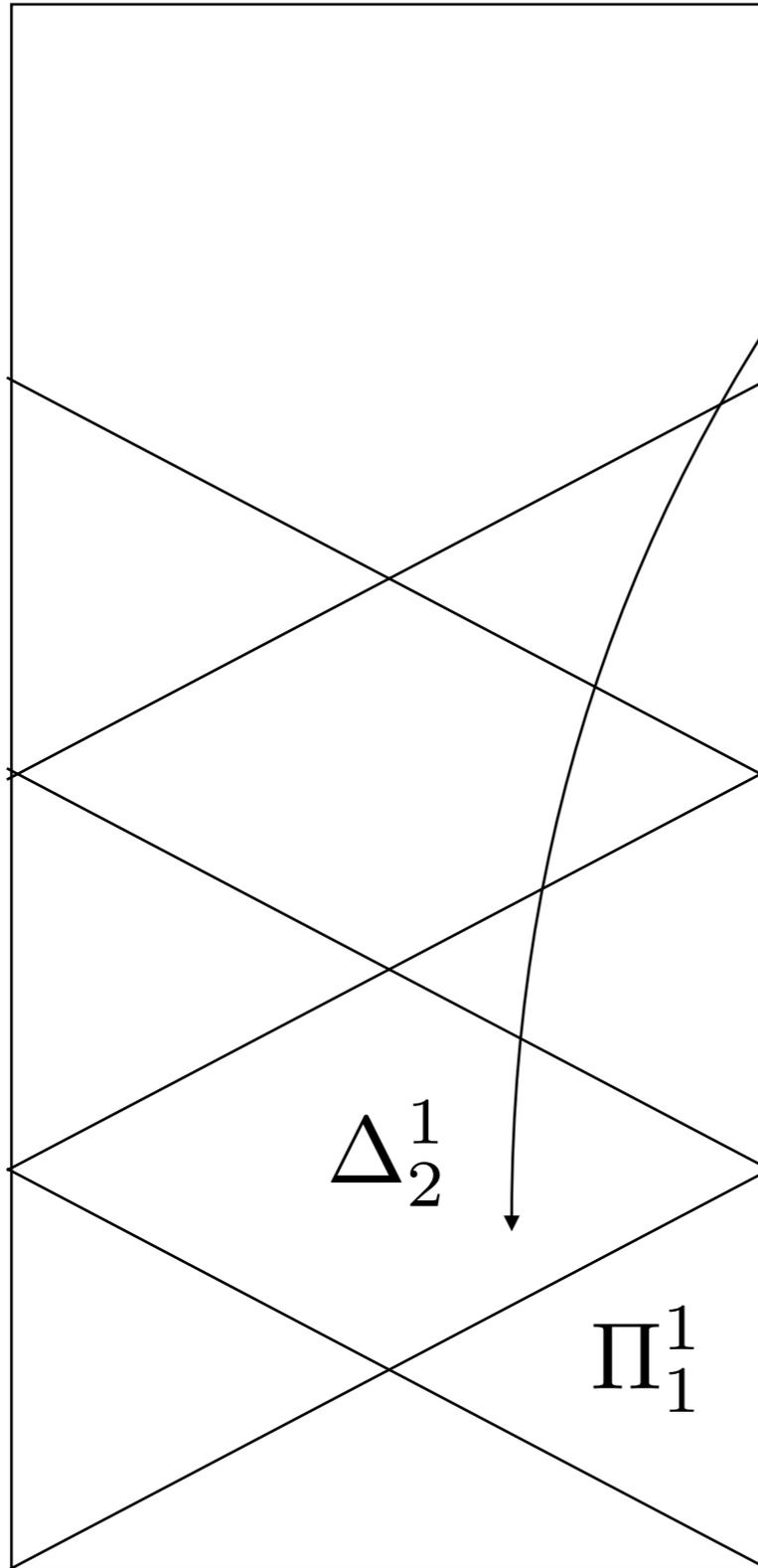
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CogSci and AI need to say more about where AI falls/can fall in the landscape.

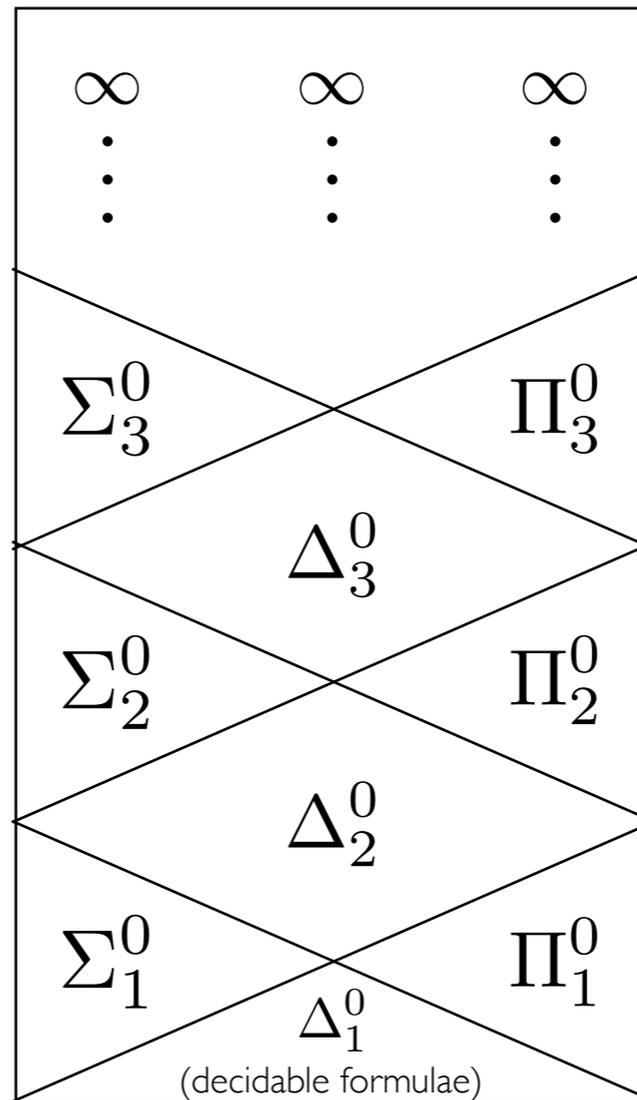


$A^n \mathcal{H}$  (Analytic Hierarchy)



Infinite Time Turing Machines (ITTMs)

$A^r \mathcal{H}$  (Arithmetic Hierarchy)

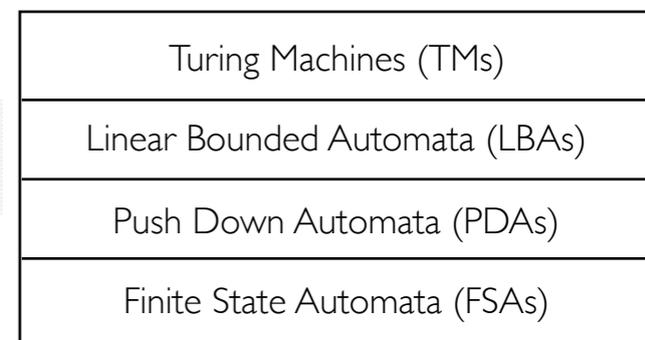


Human Persons (according to Bringsjord)

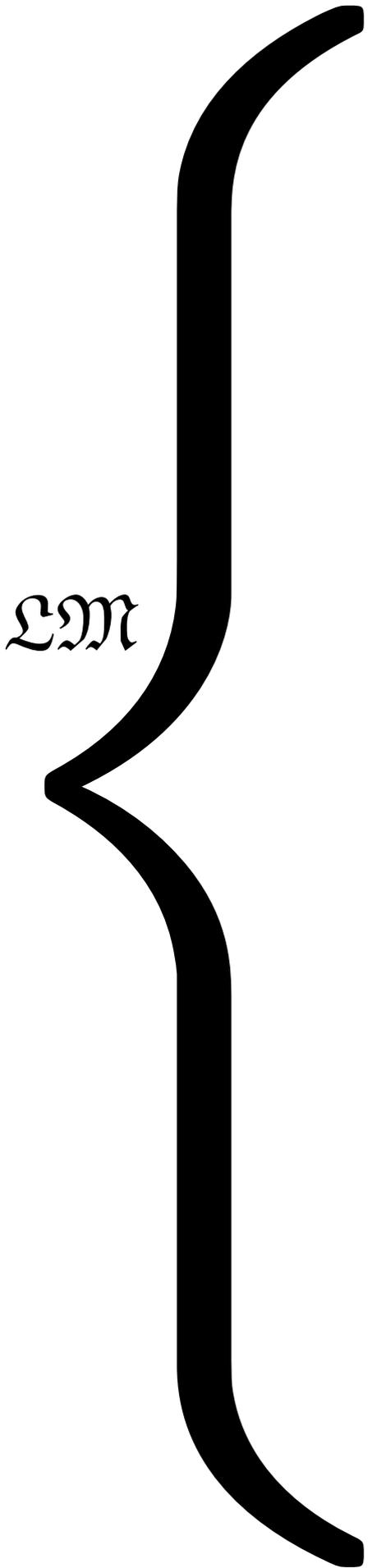
Human Brains (according to Granger)



$\mathcal{CH}$  (Chomsky Hierarchy)

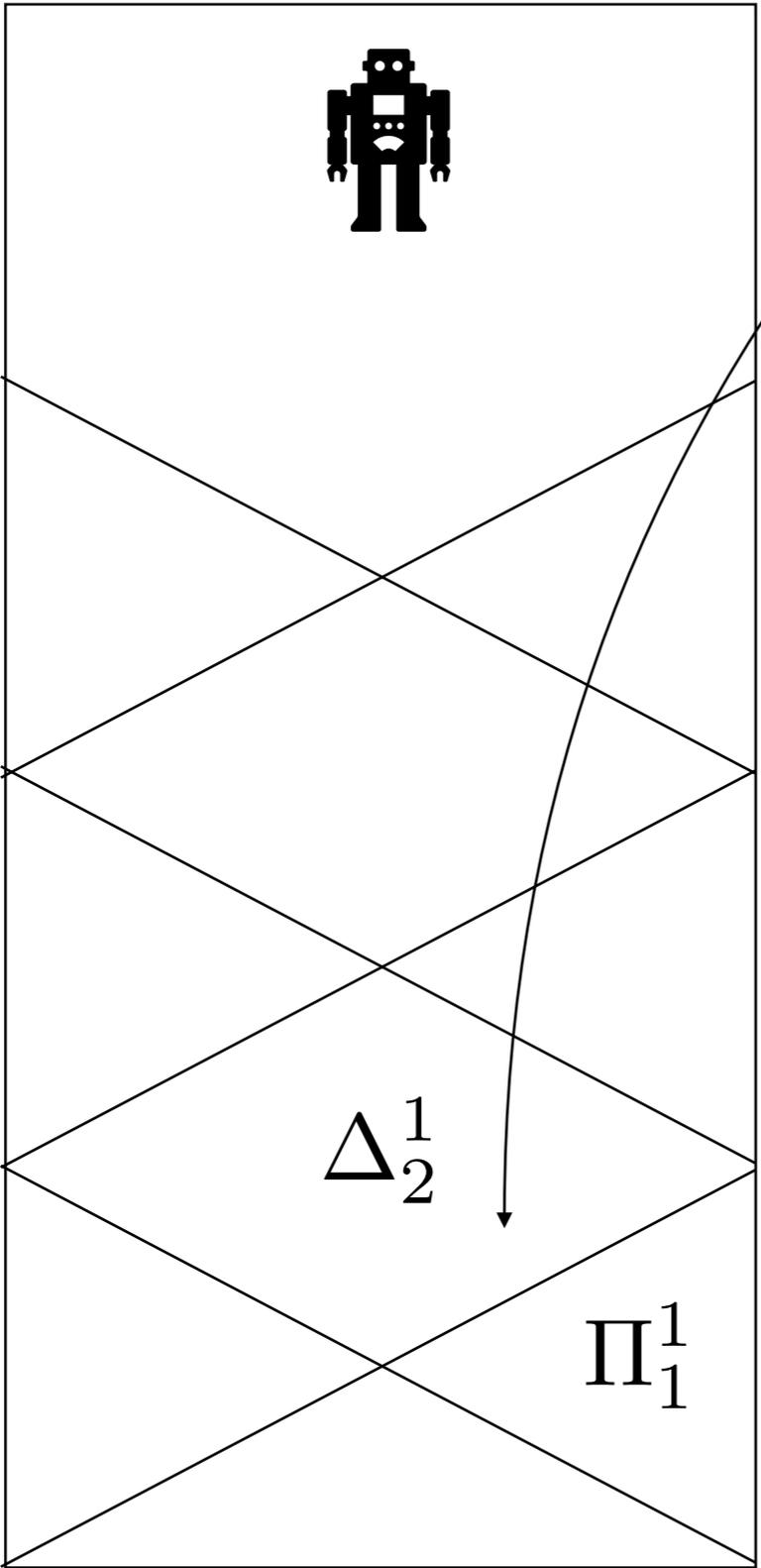


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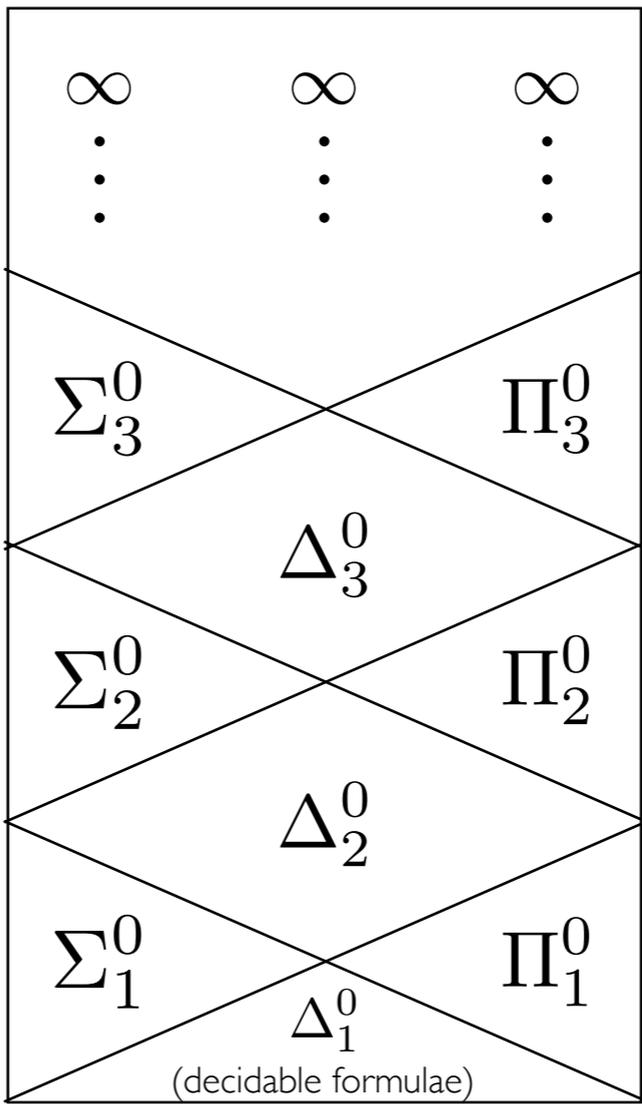
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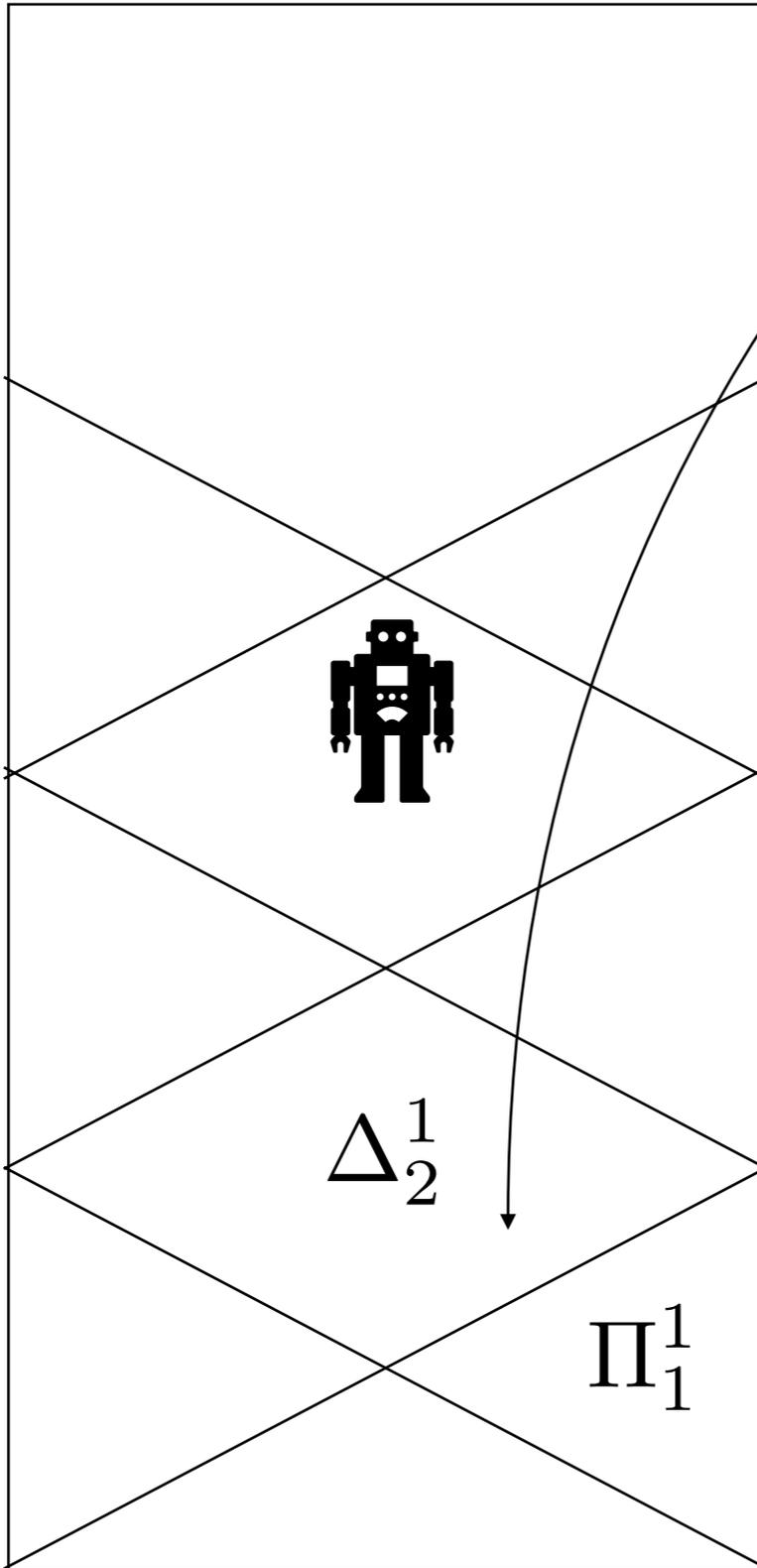
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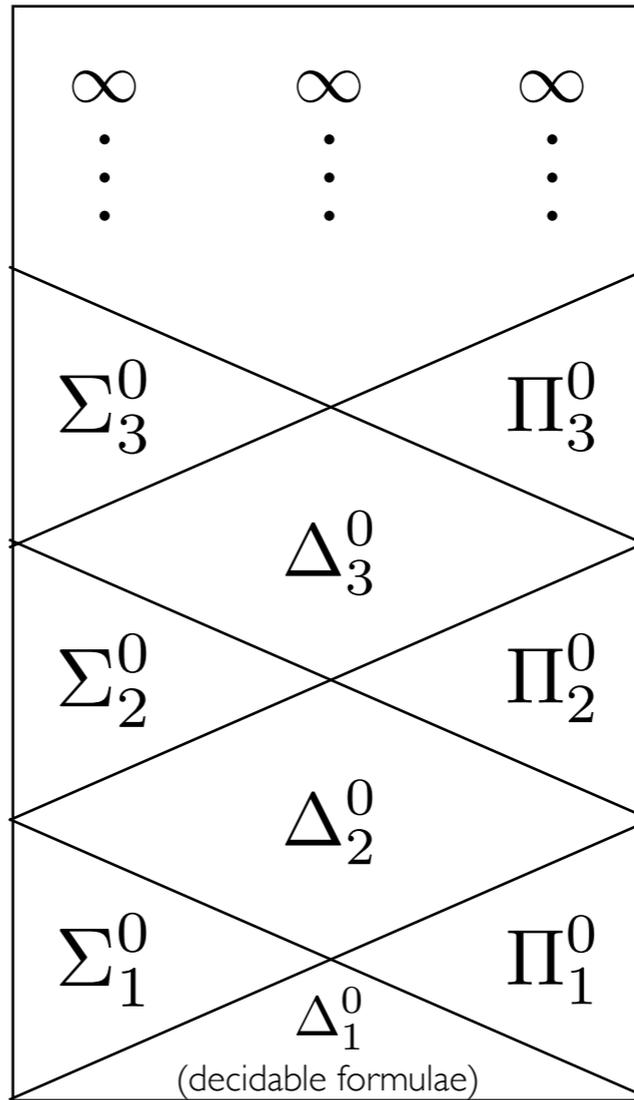
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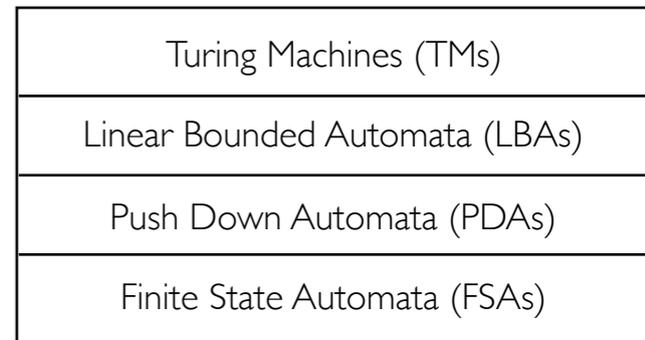
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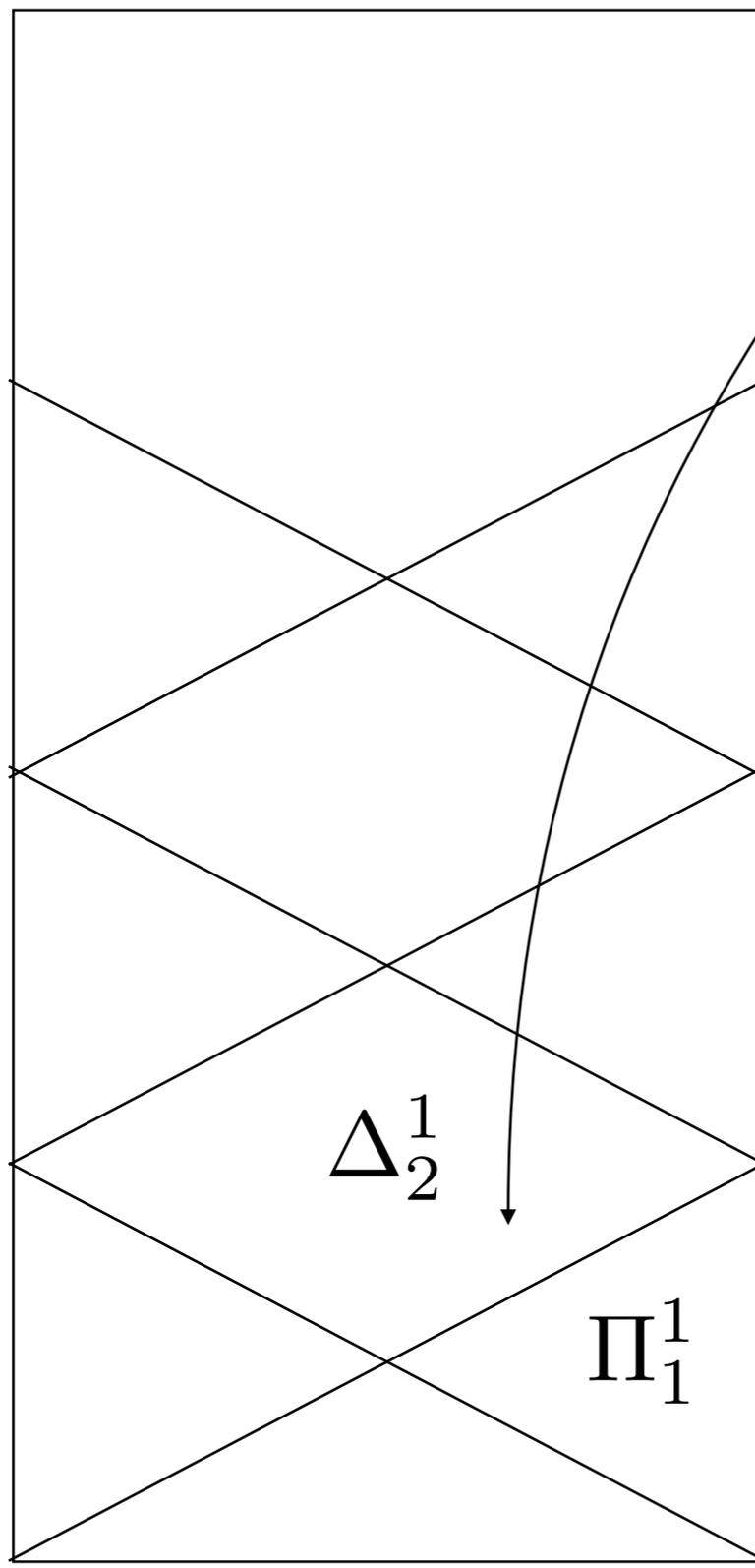


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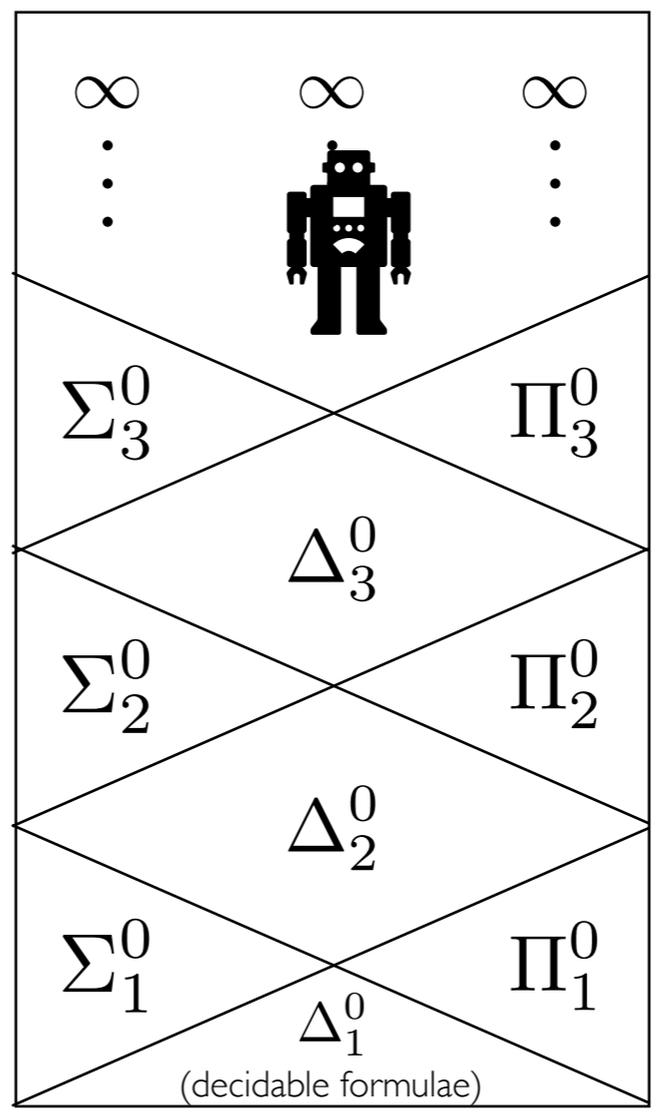
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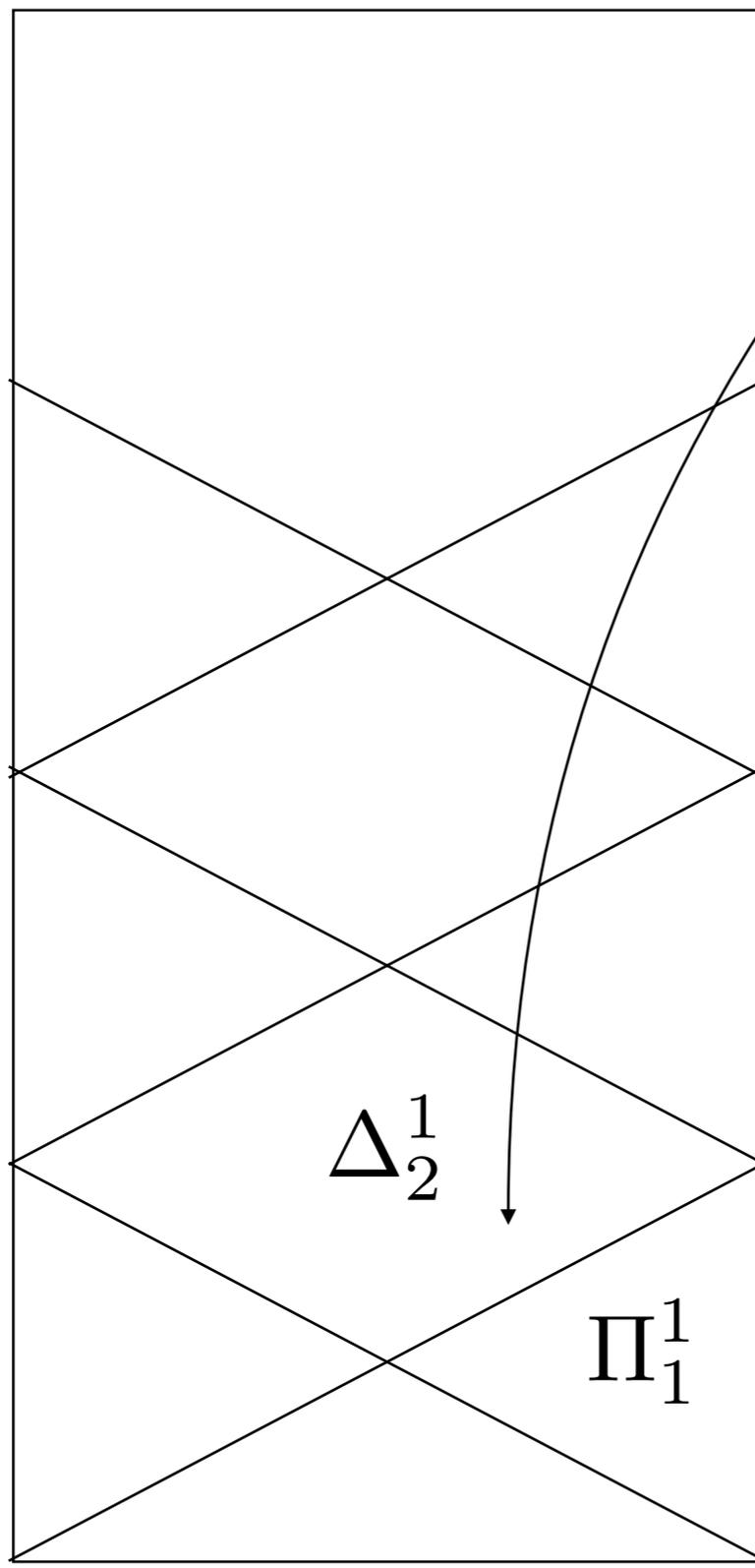
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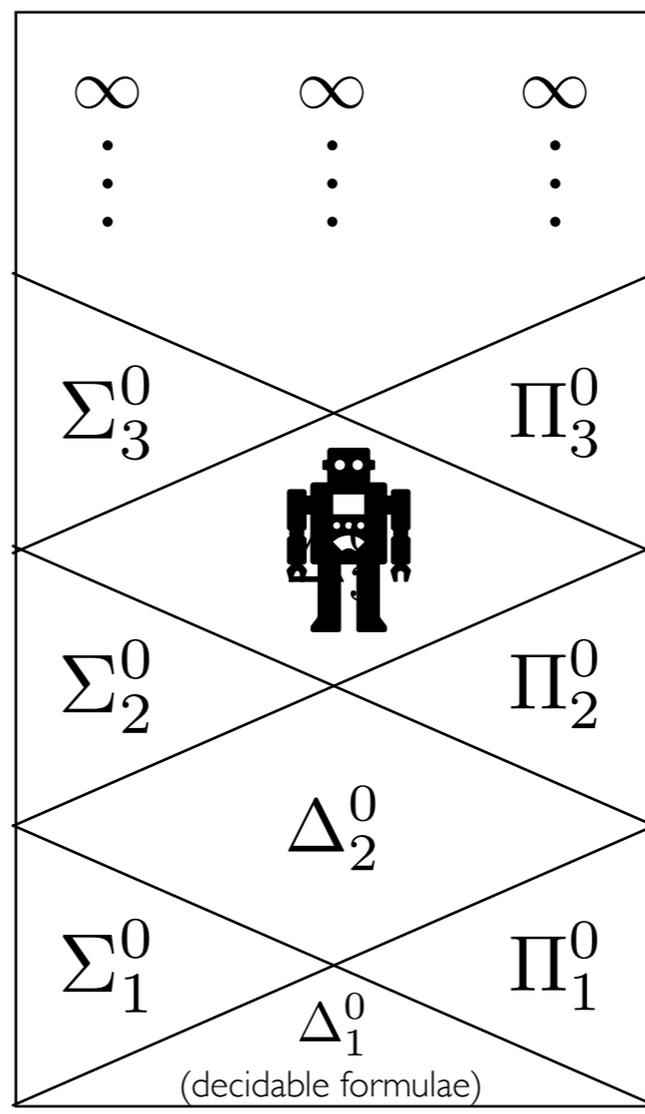
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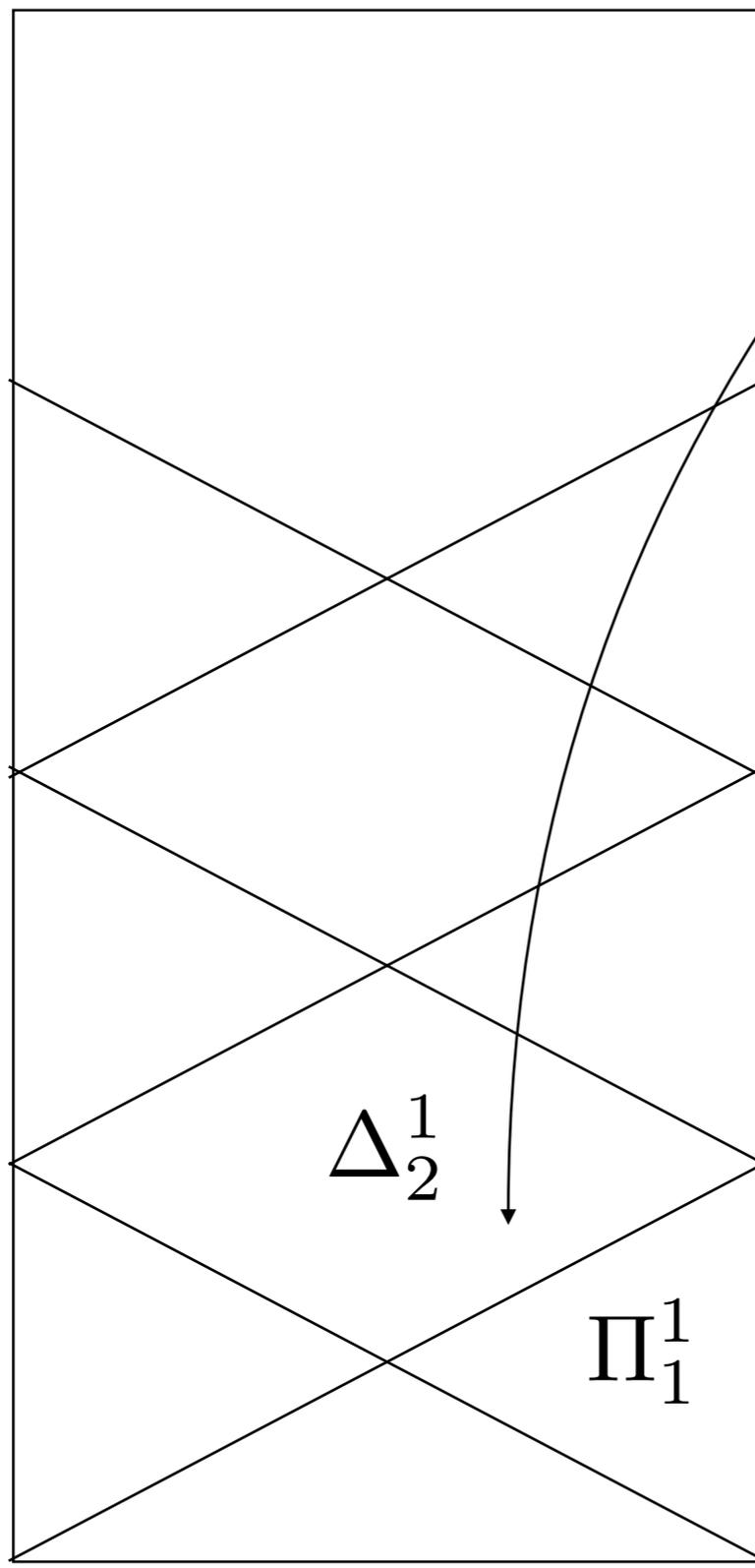
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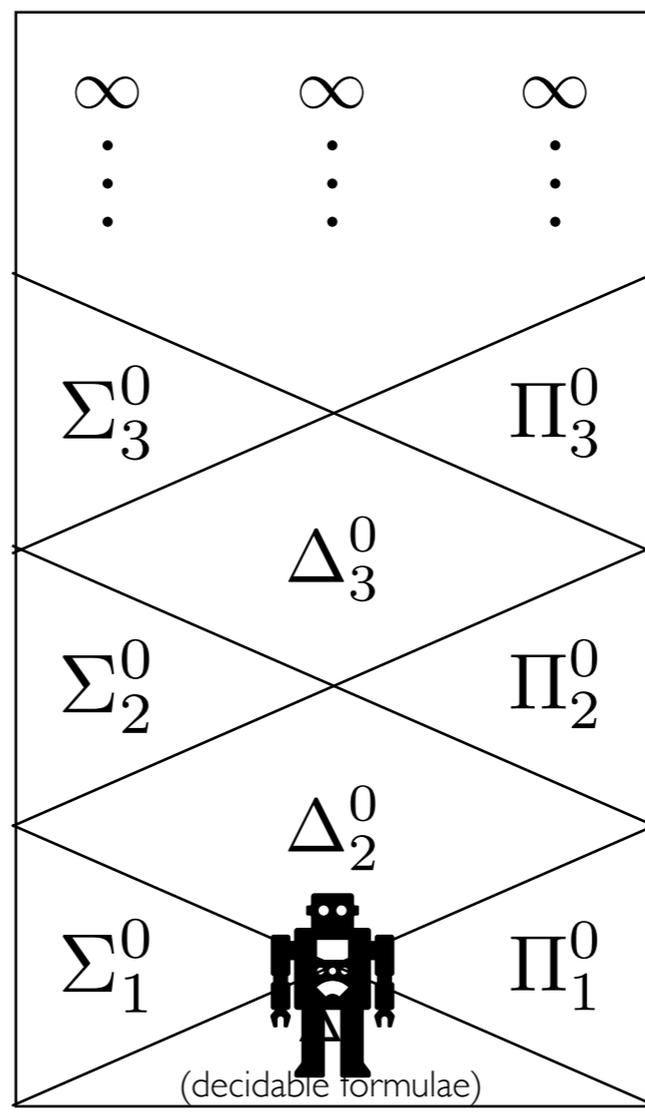
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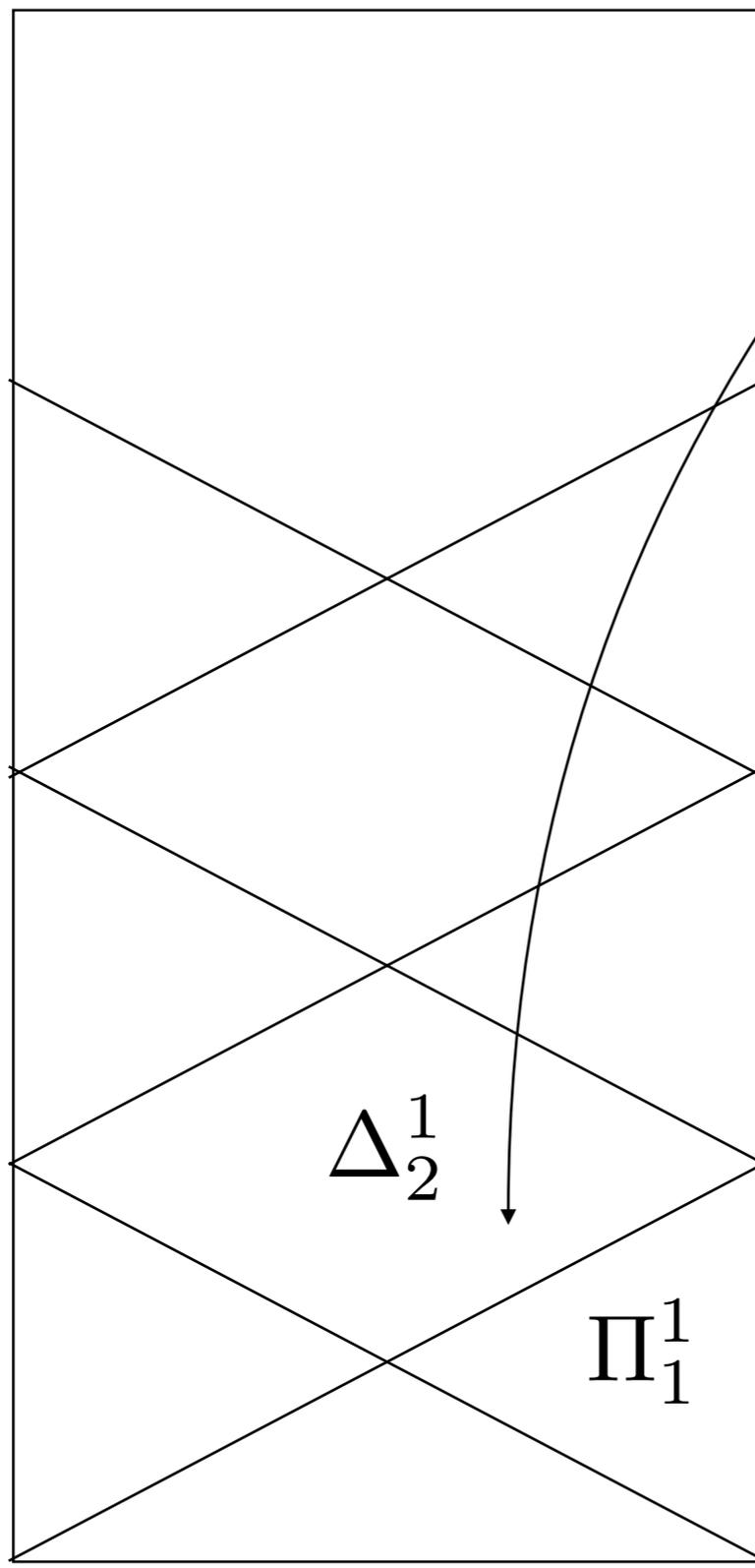
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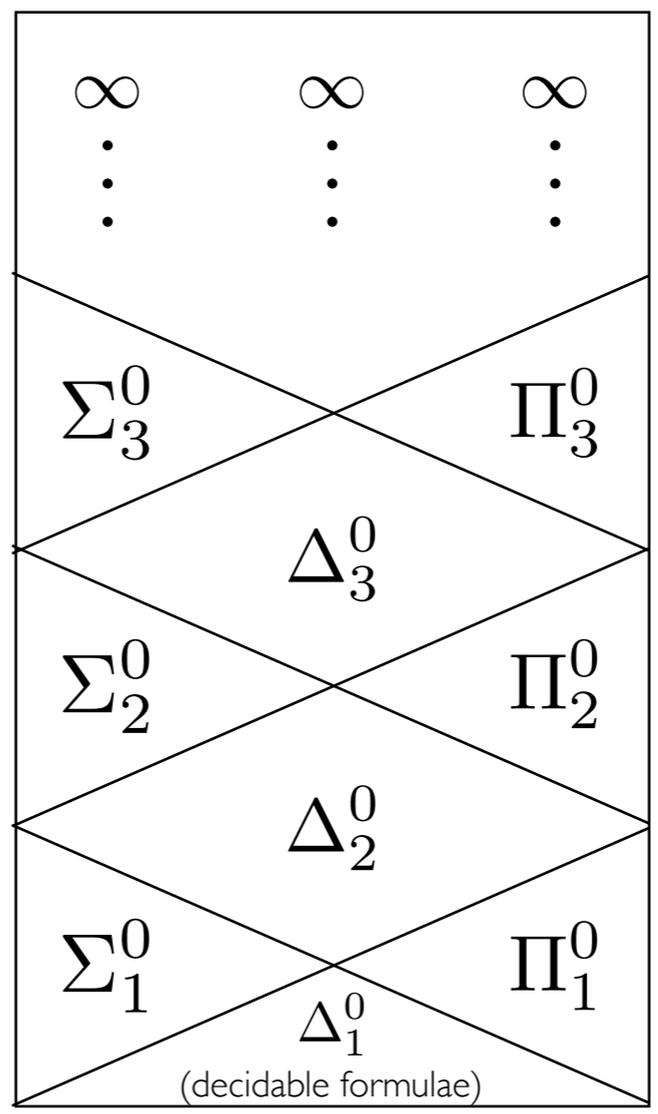
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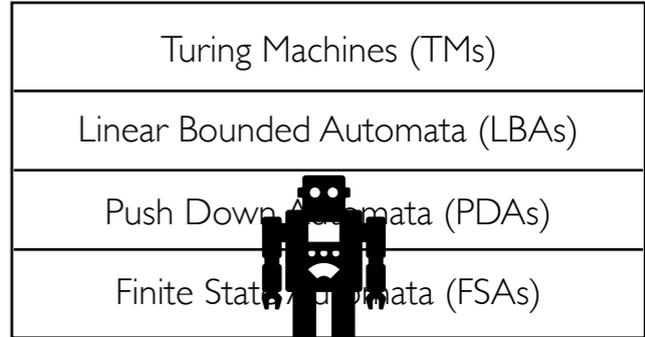
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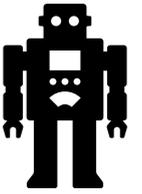


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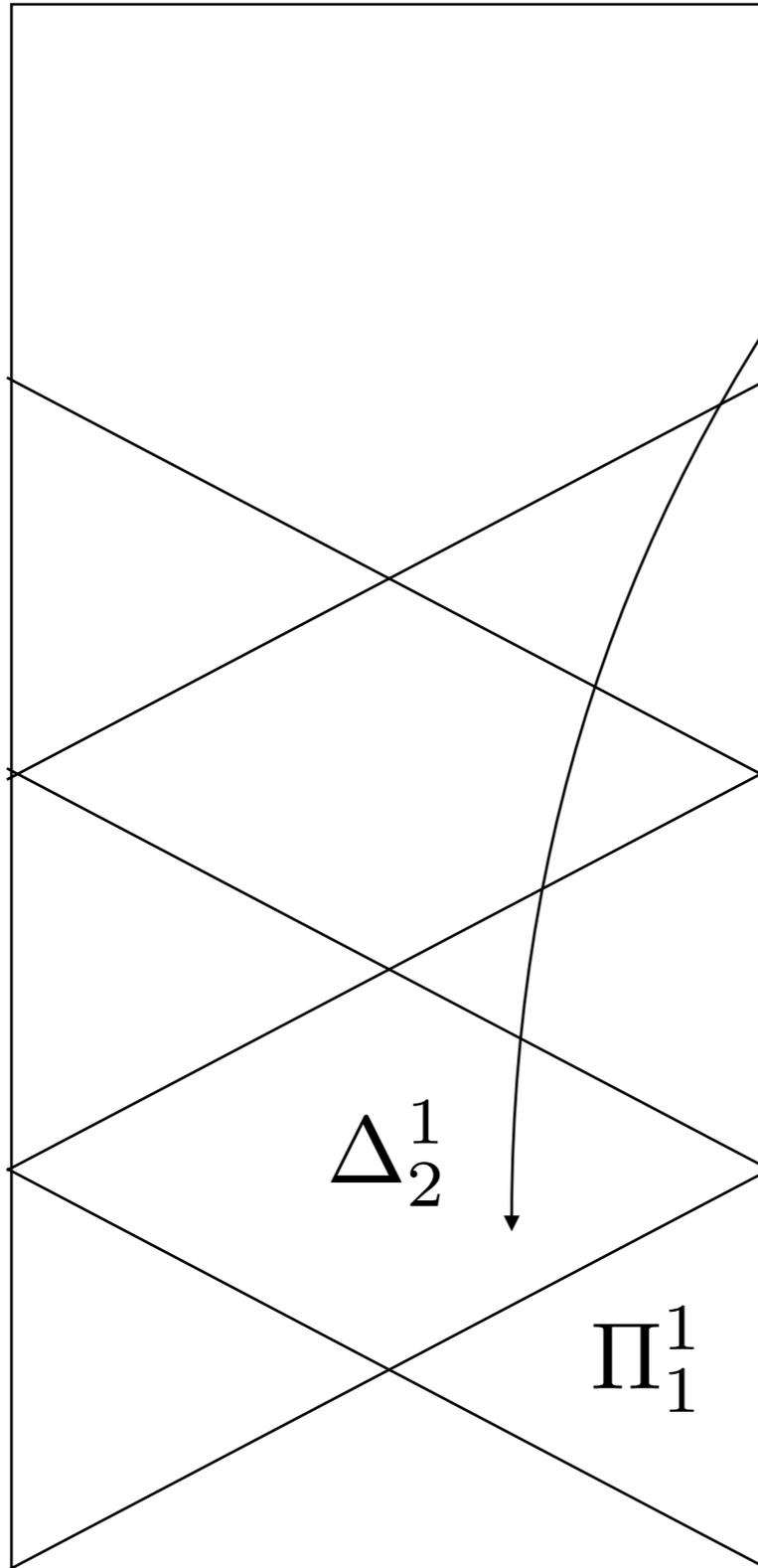


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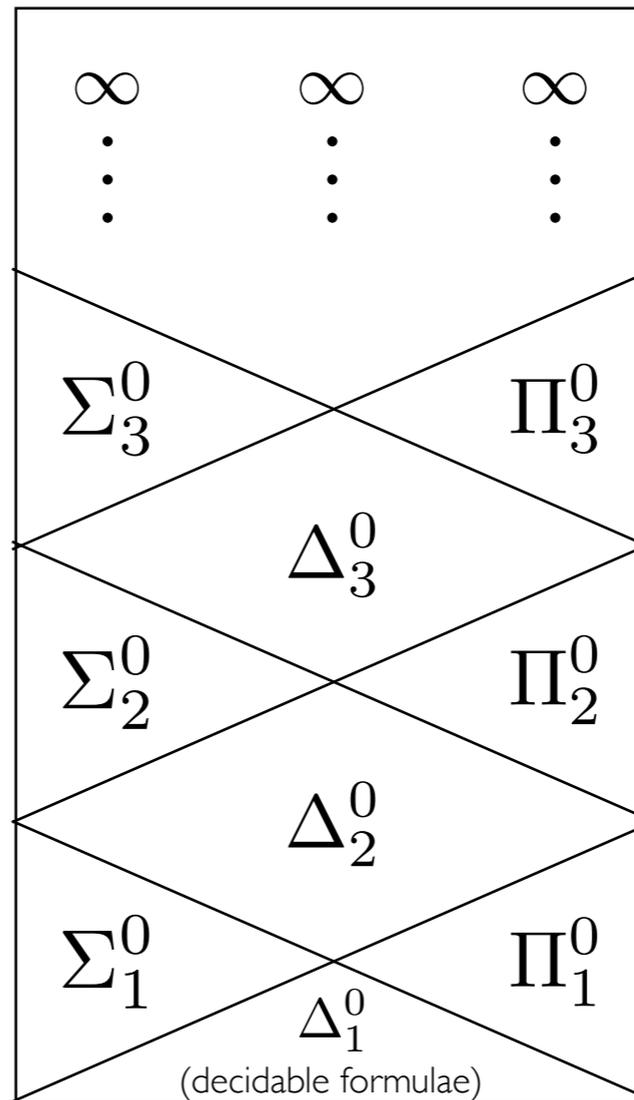


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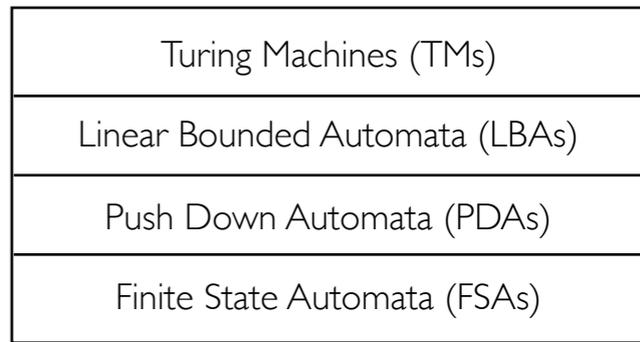


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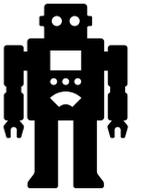


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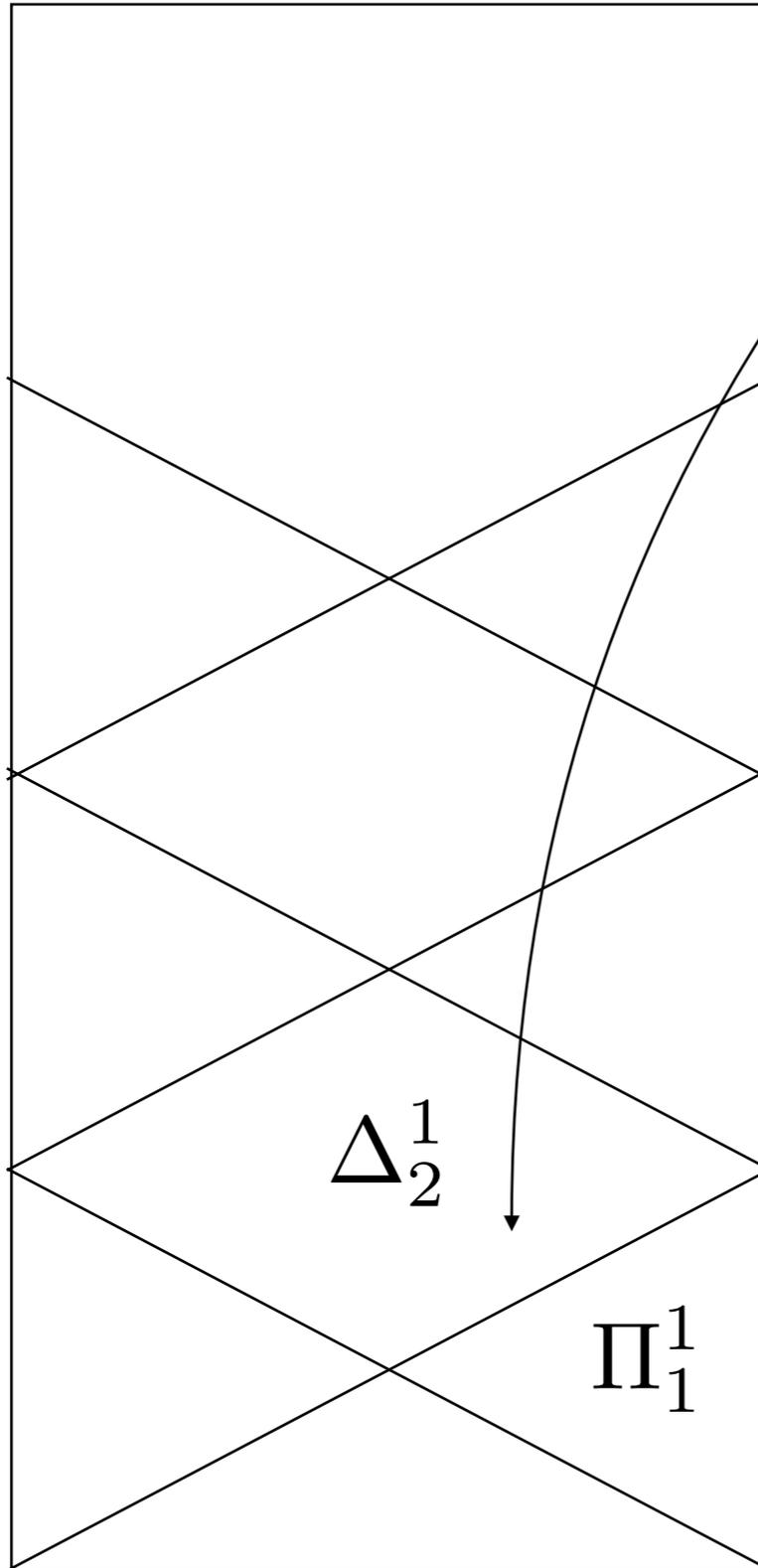


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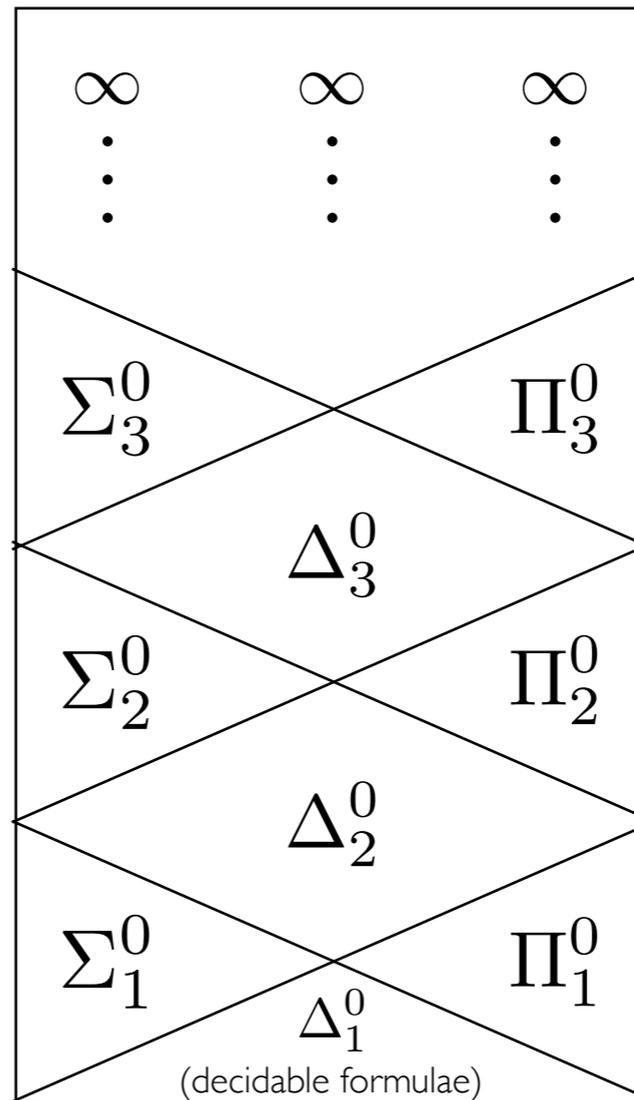


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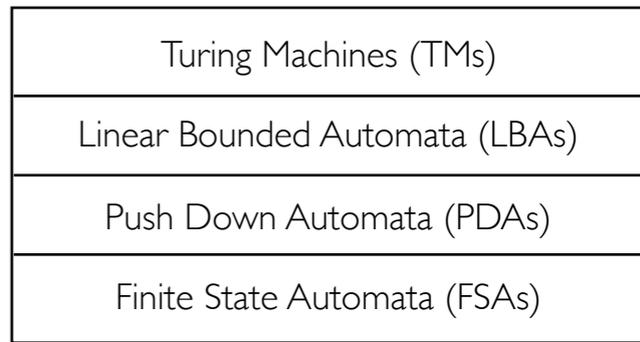


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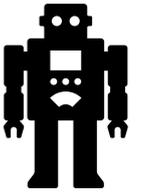


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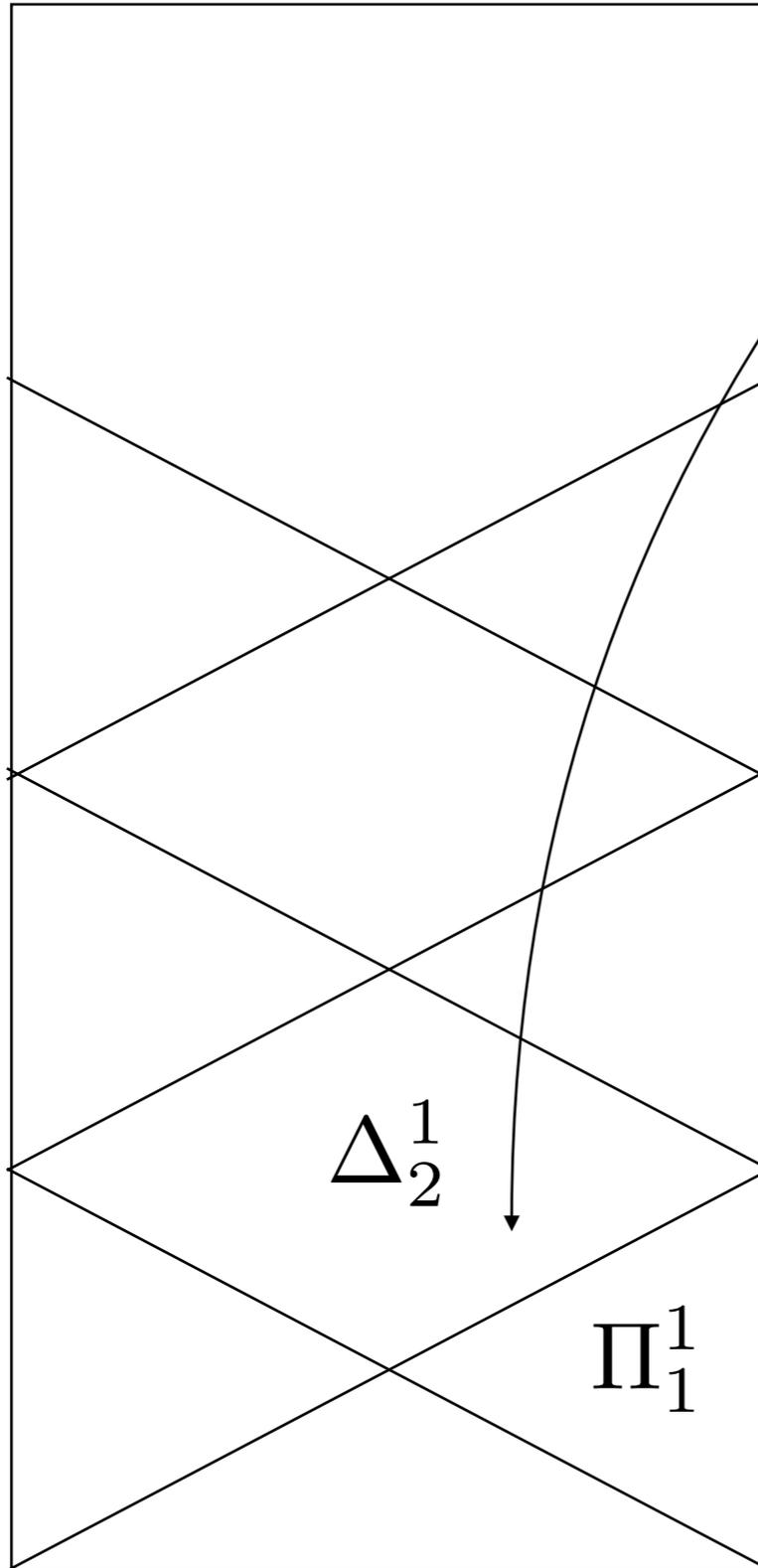


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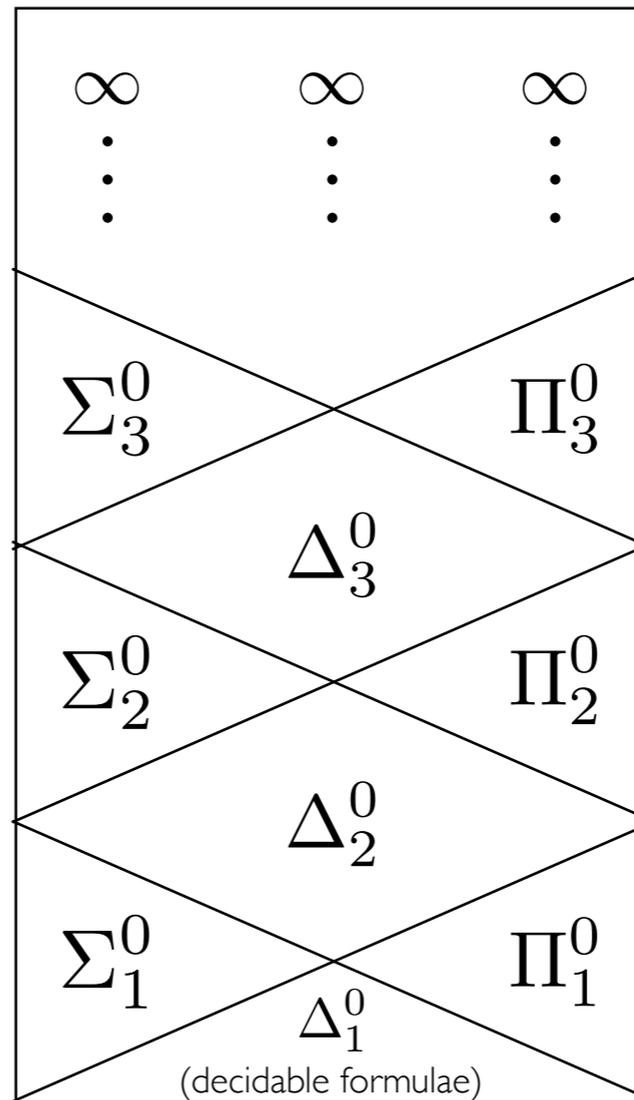


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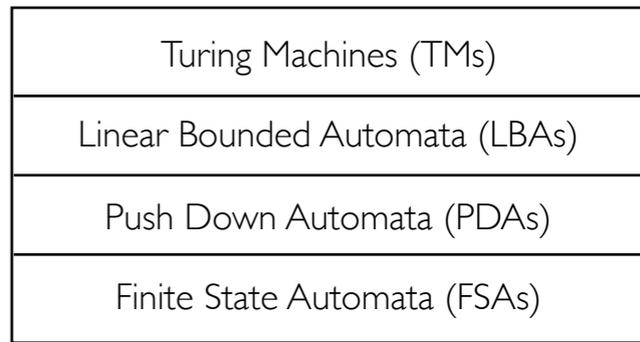


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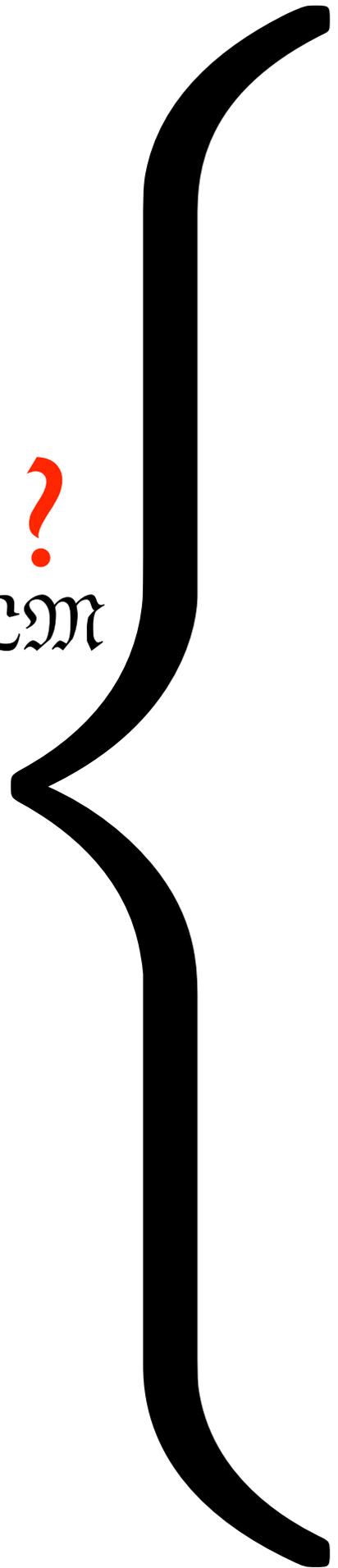
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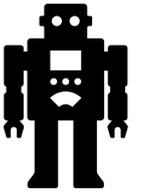
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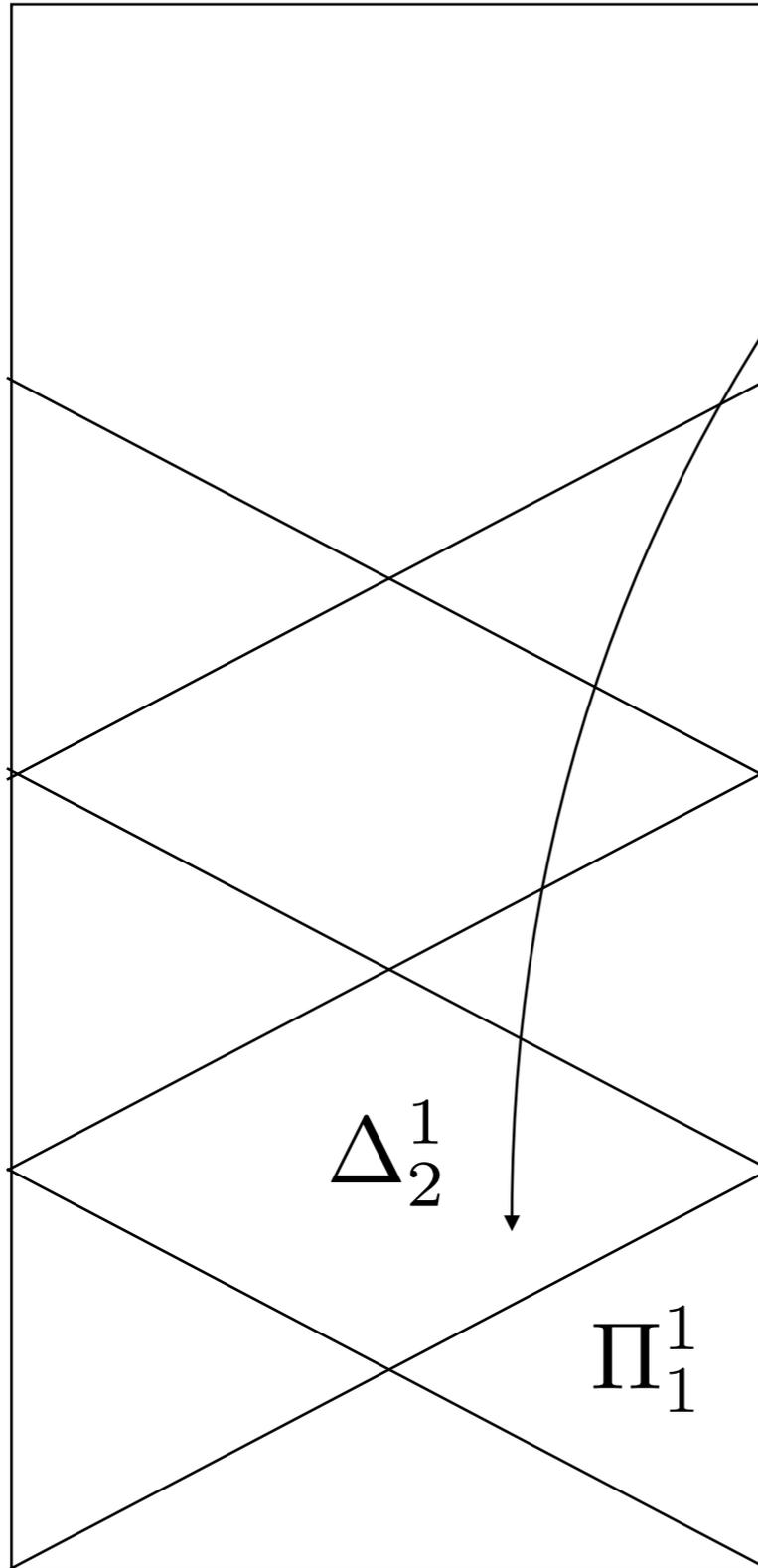
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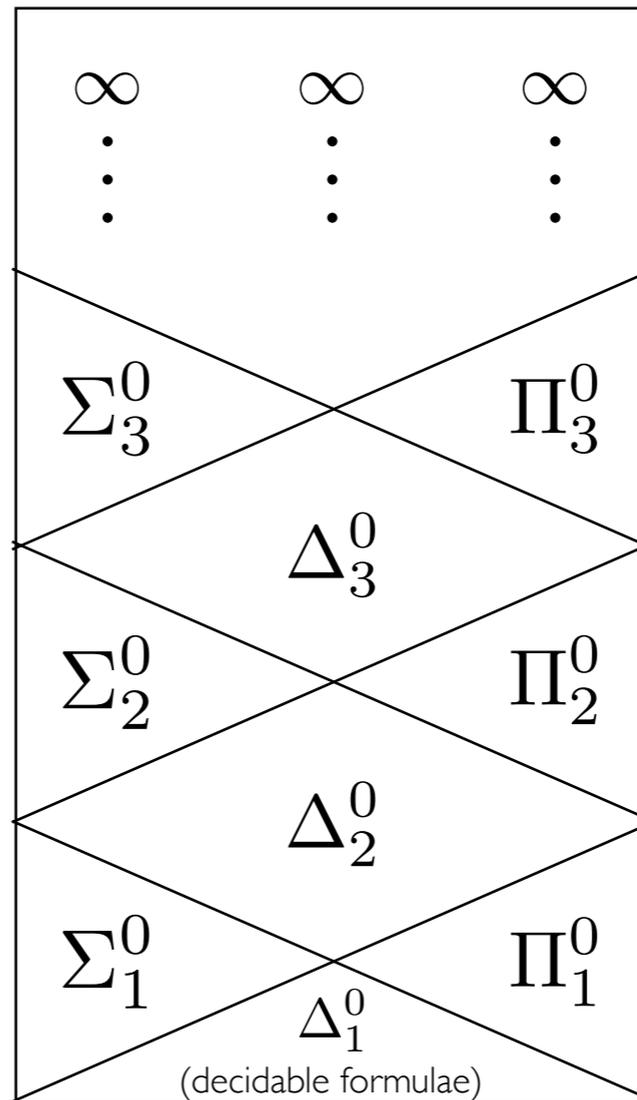


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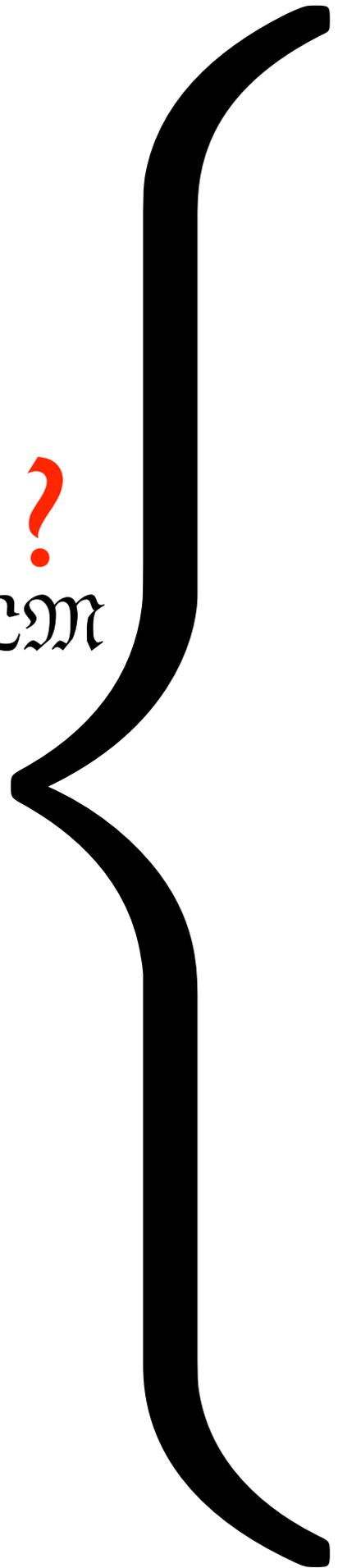
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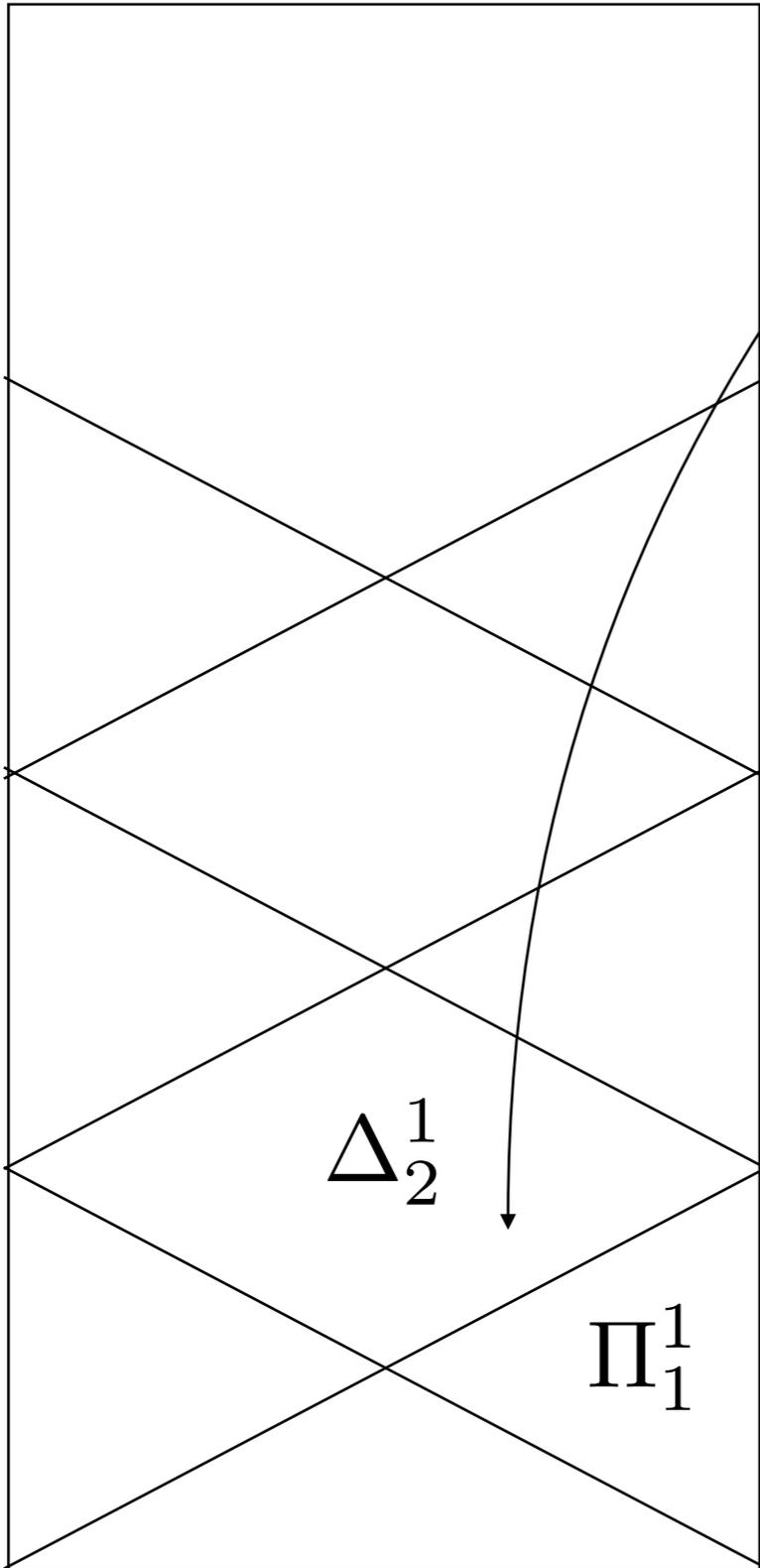
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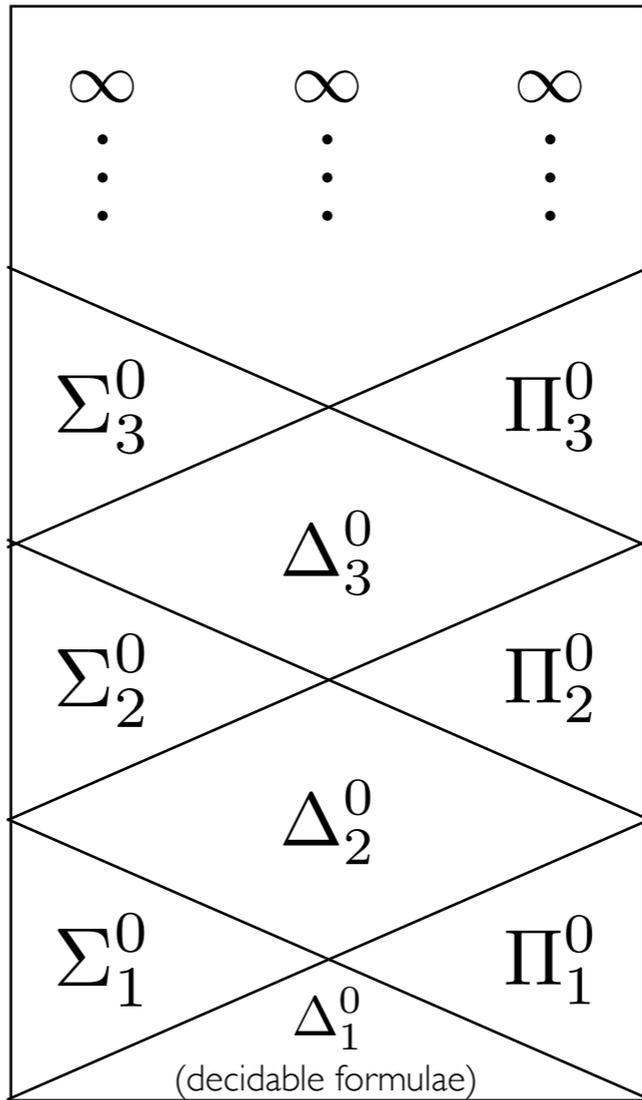
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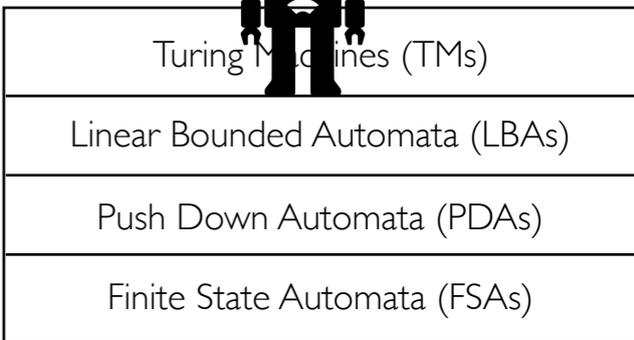
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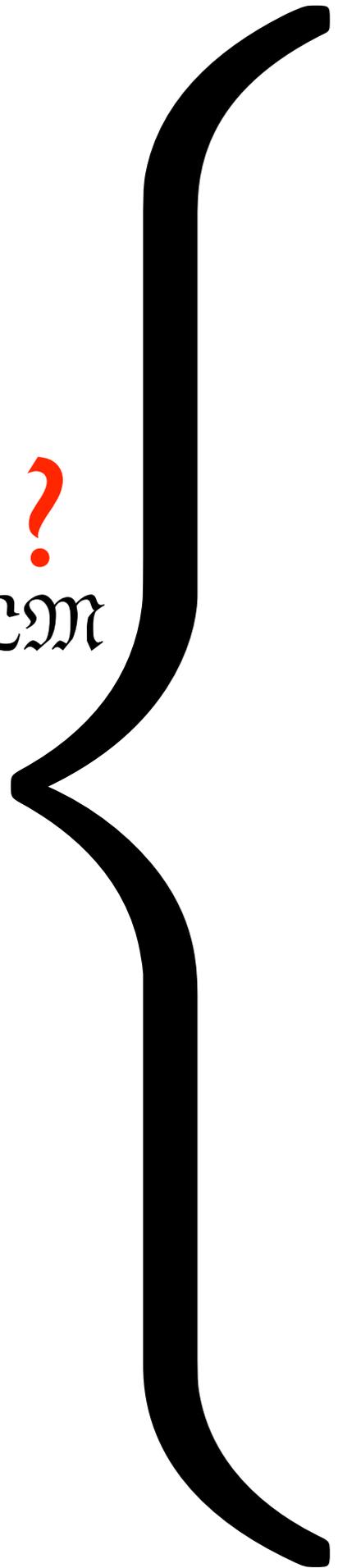
Human Brains (according to Granger)



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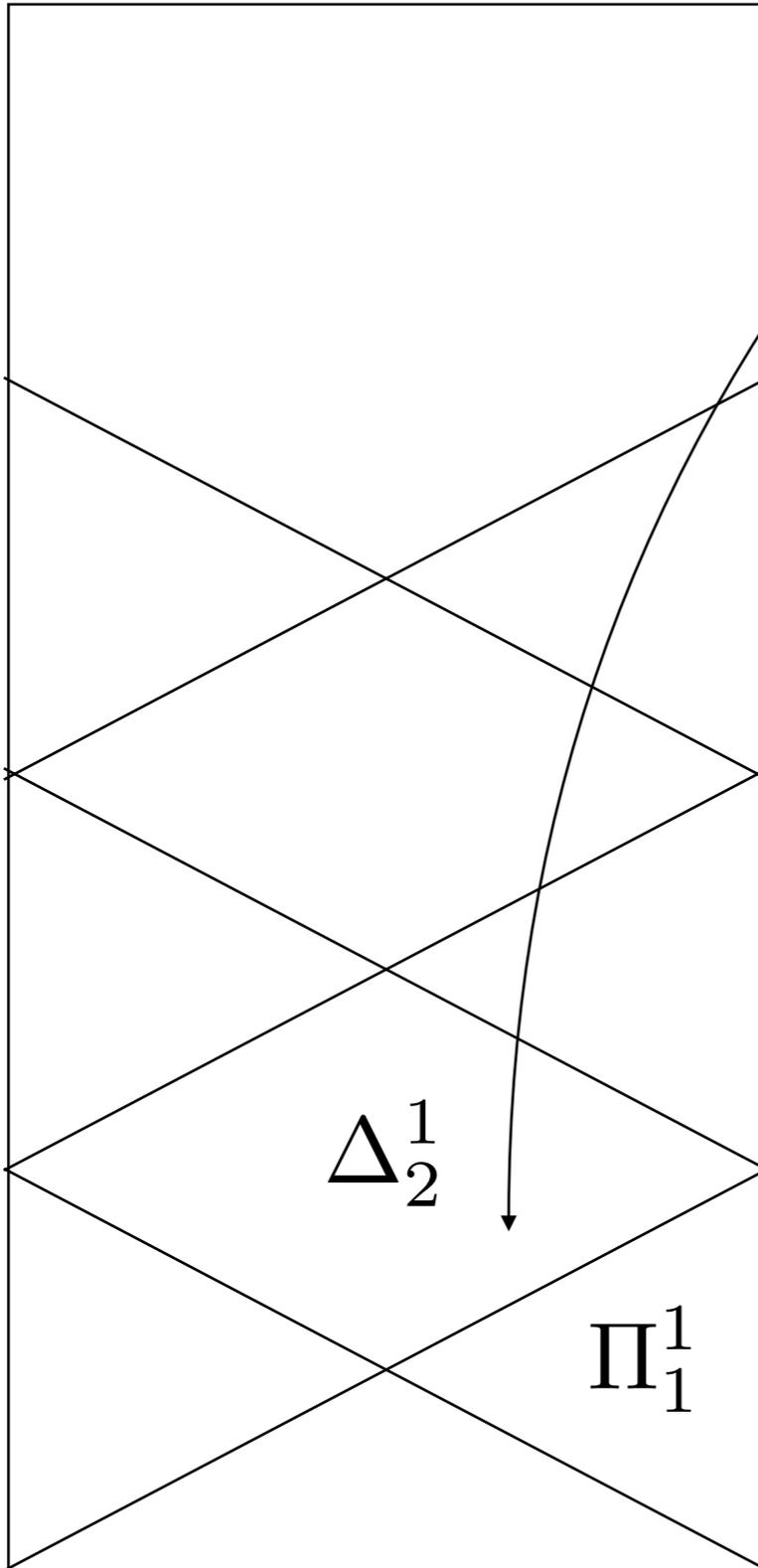


?  
 $\mathcal{EM}$



CogSci and AI need to say more about where AI falls/can fall in the landscape.

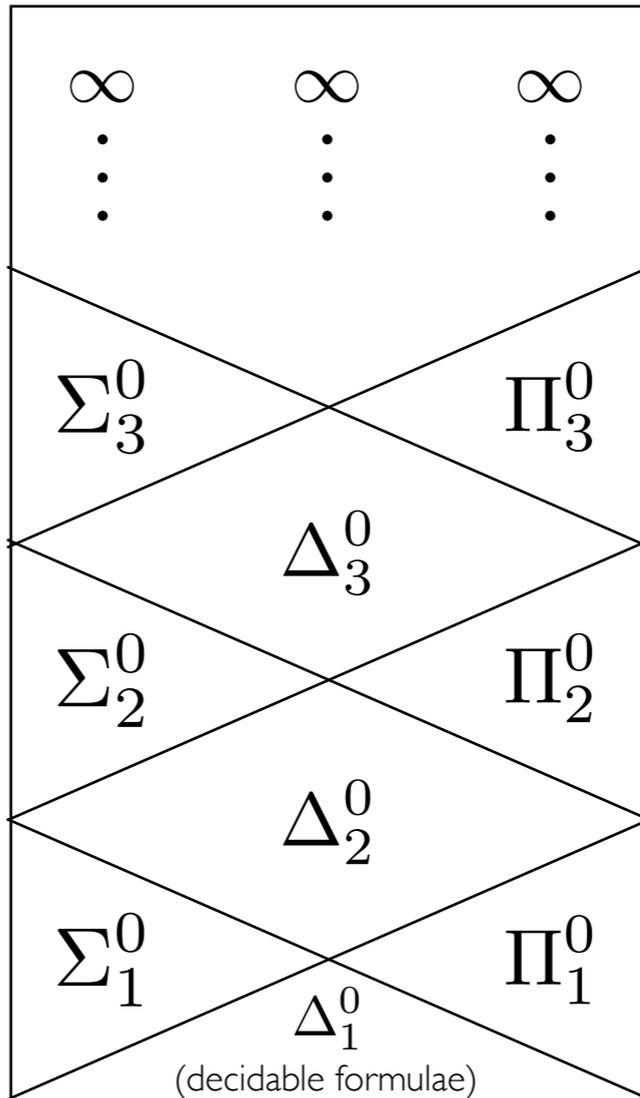
$A^n \mathcal{H}$  (Analytic Hierarchy)



Infinite Time Turing Machines (ITTMs)

Human Persons (according to Bringsjord)

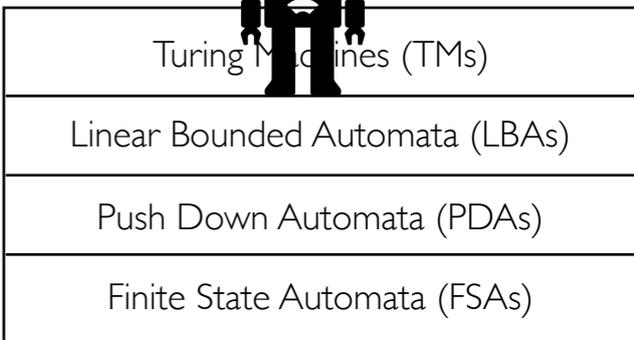
$A^r \mathcal{H}$  (Arithmetic Hierarchy)



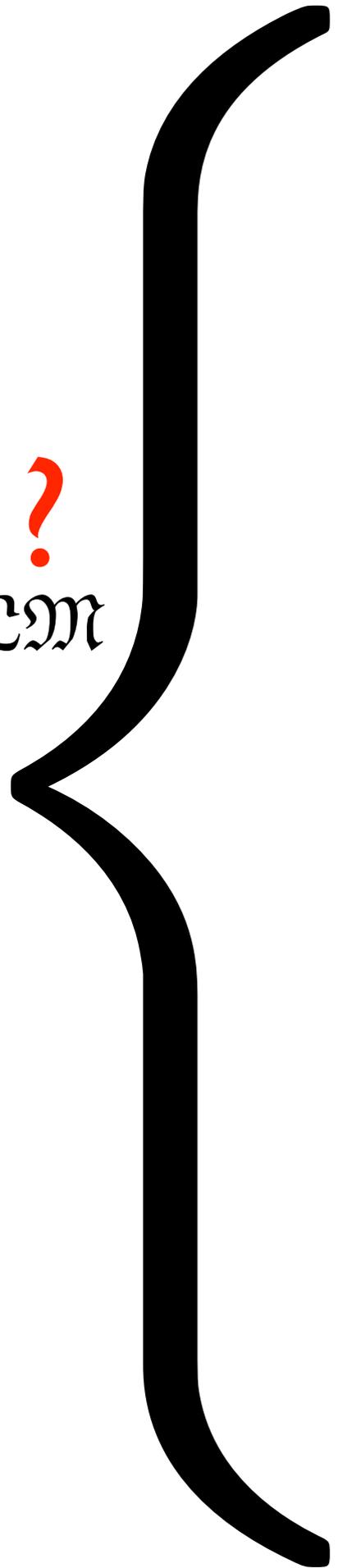
Human Brains (according to Granger)



$\mathcal{CH}$  (Chomsky Hierarchy)



?  
EM



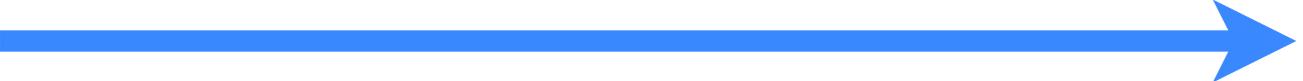




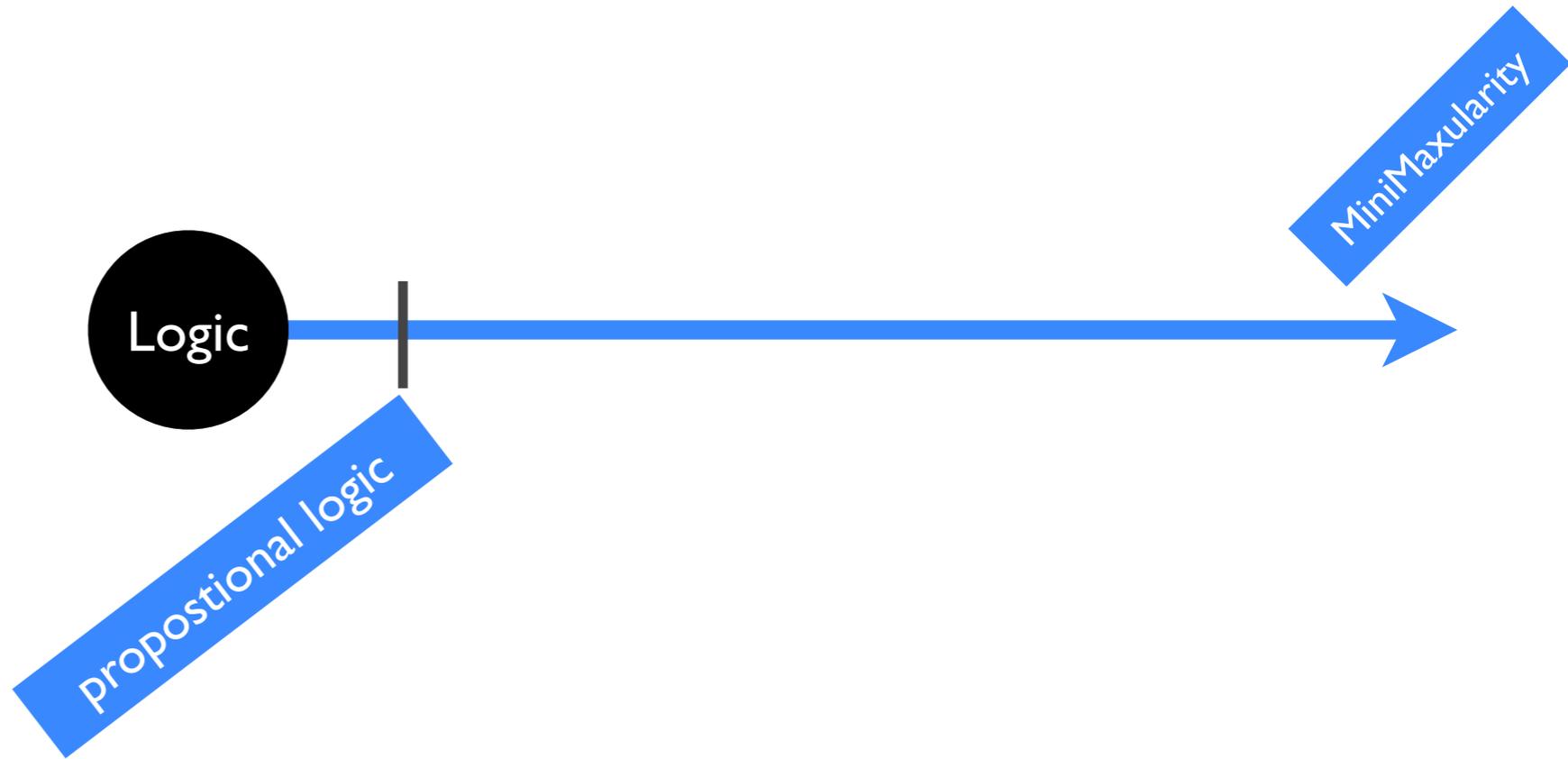
Logic

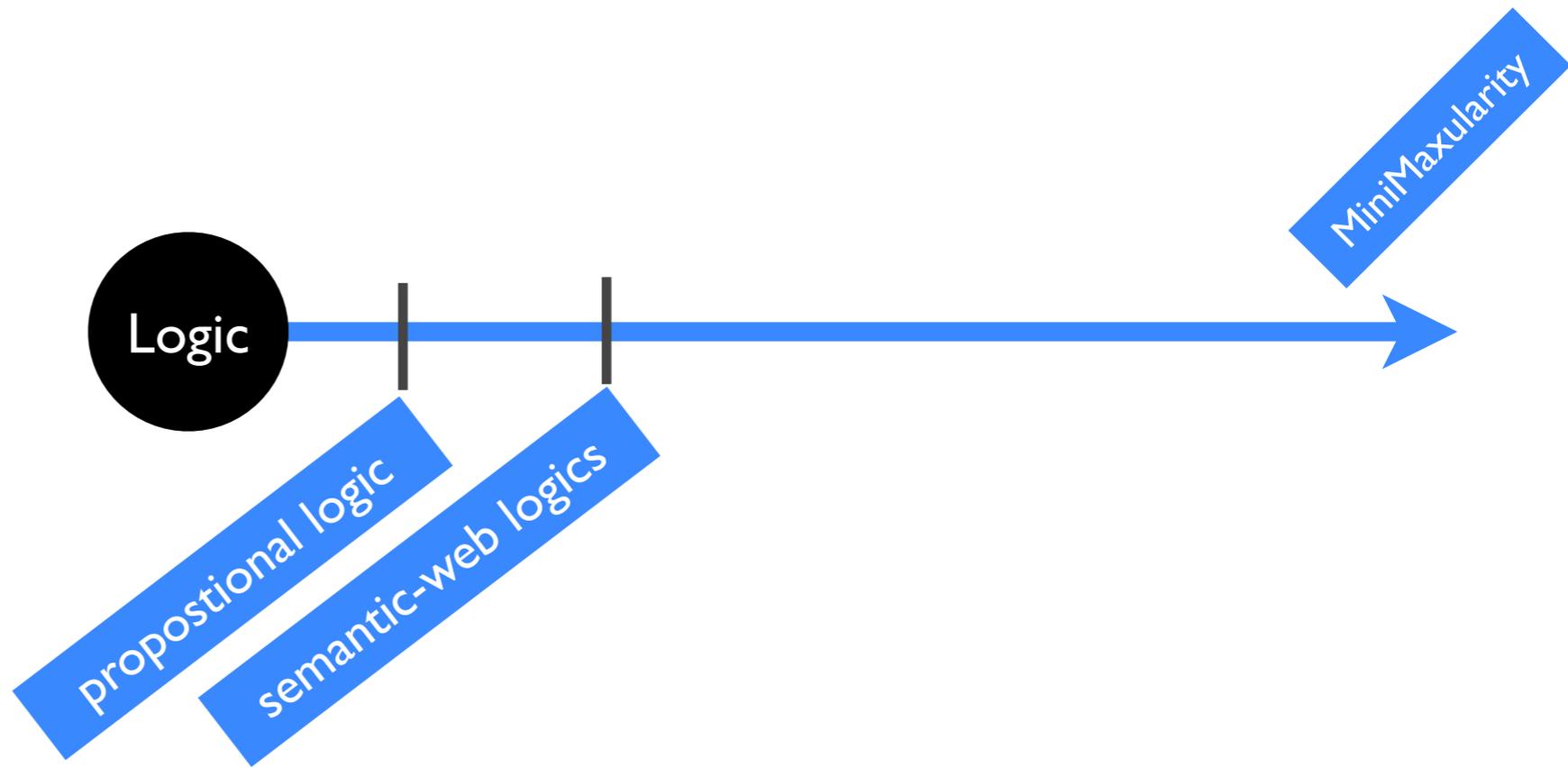


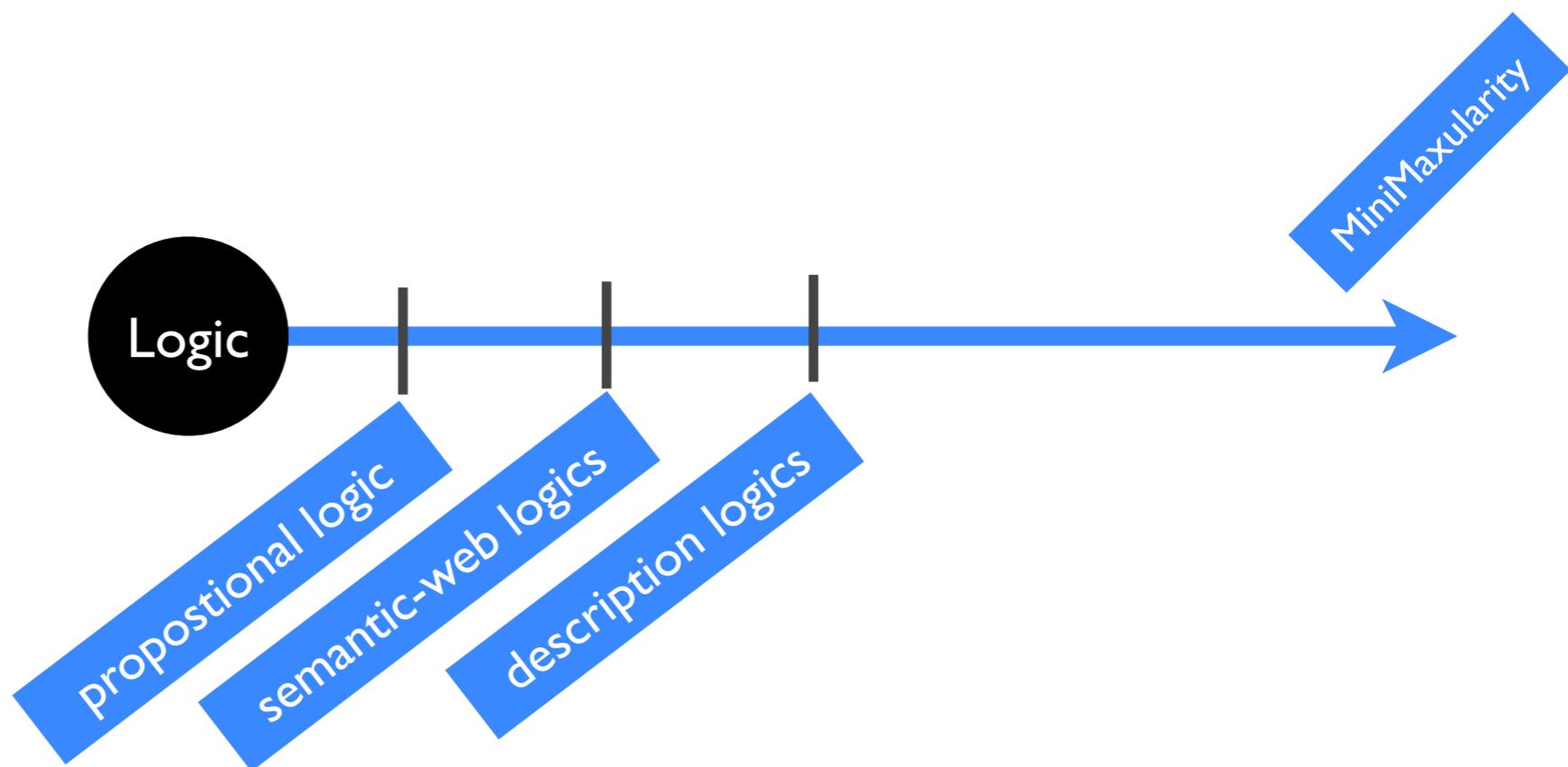
Logic

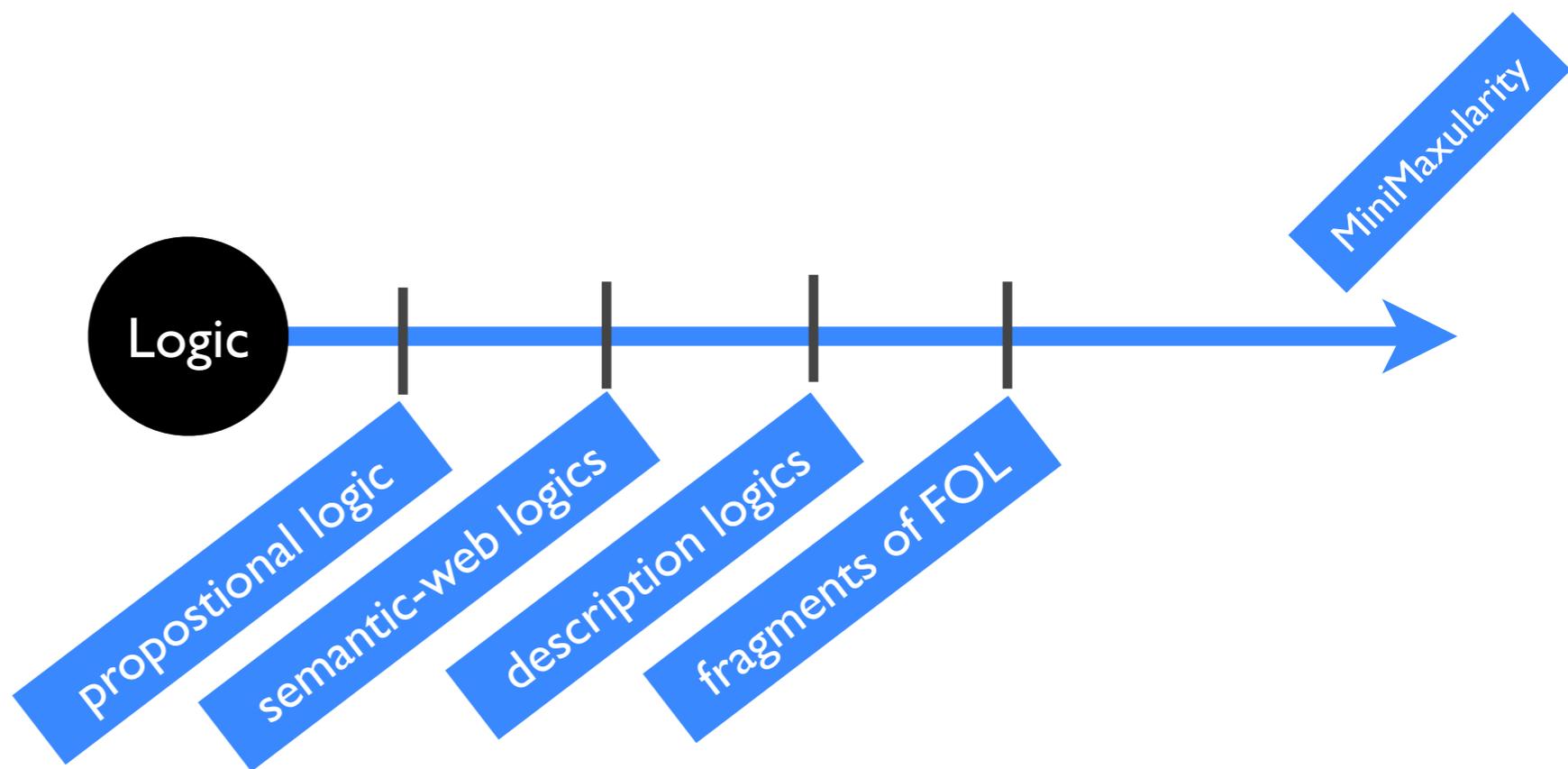


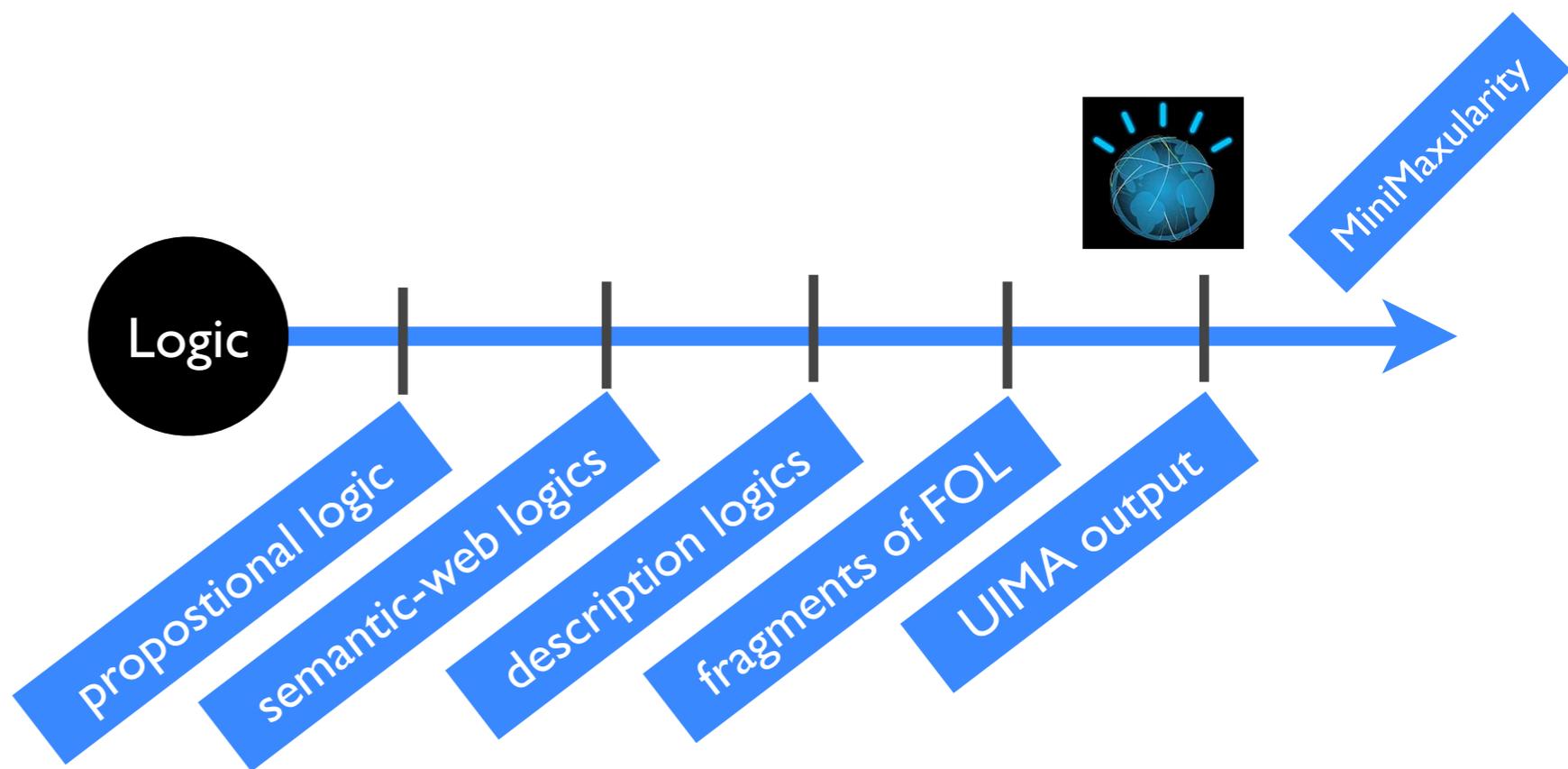
MiniMaxularity

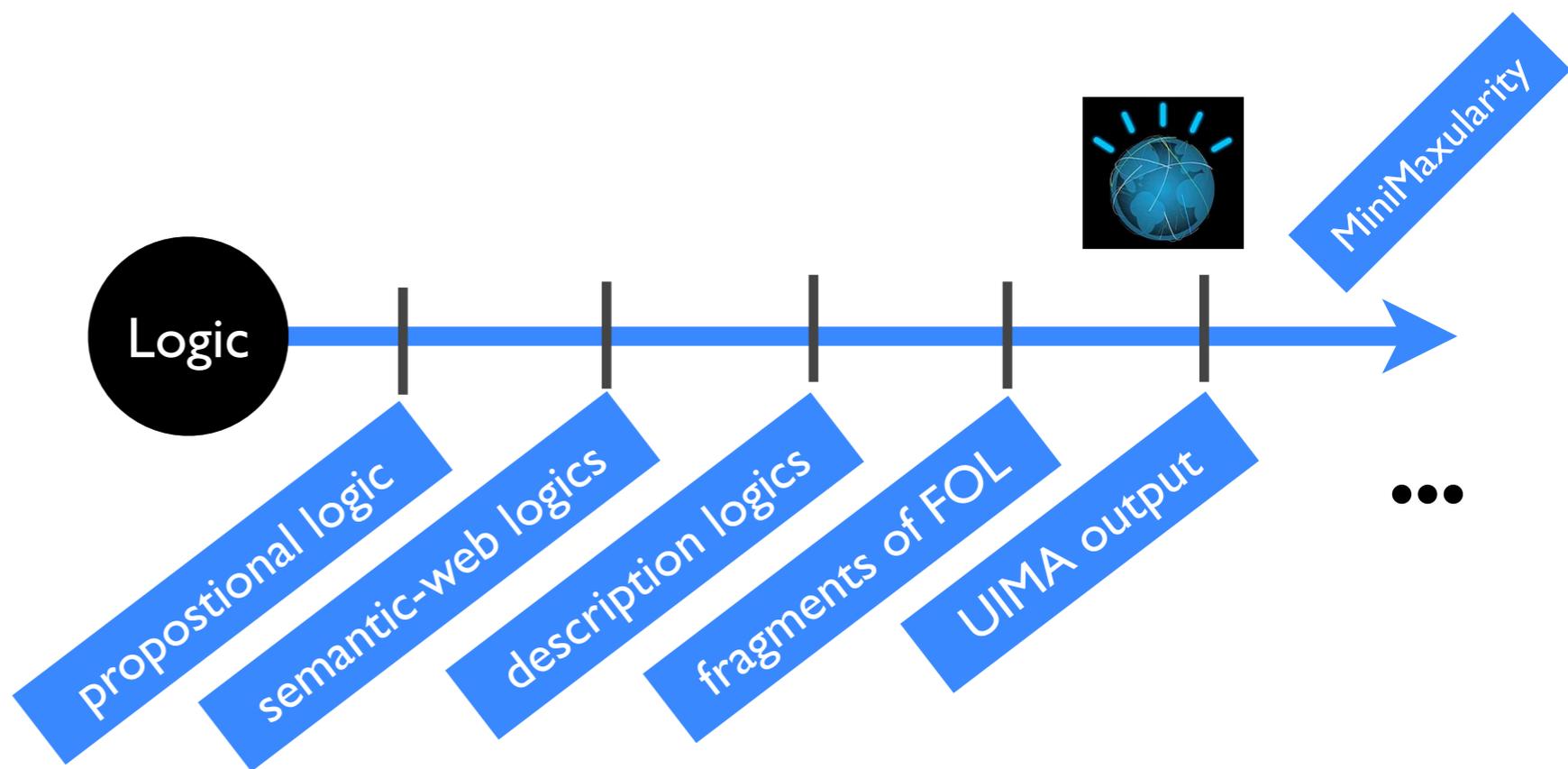


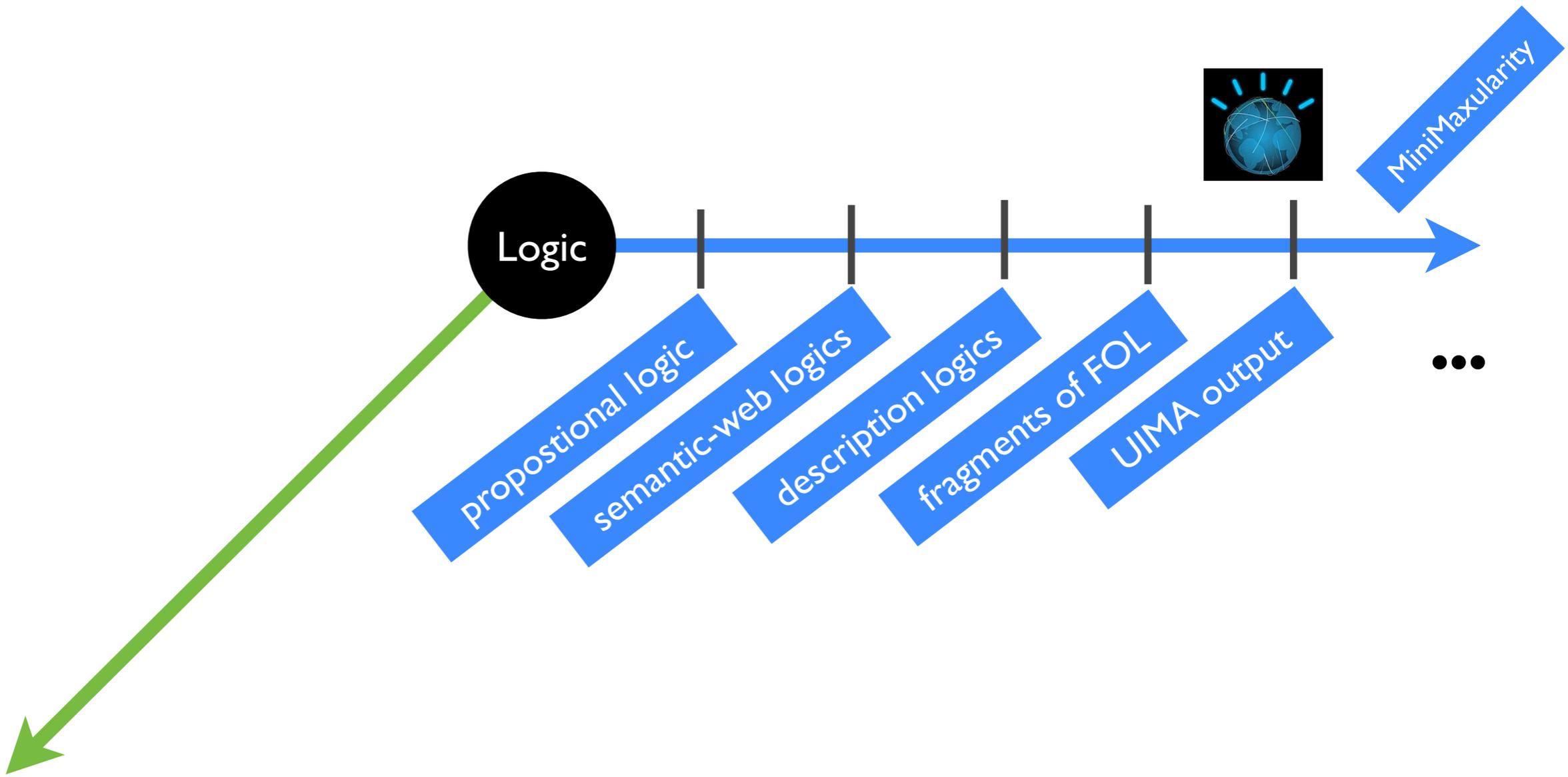






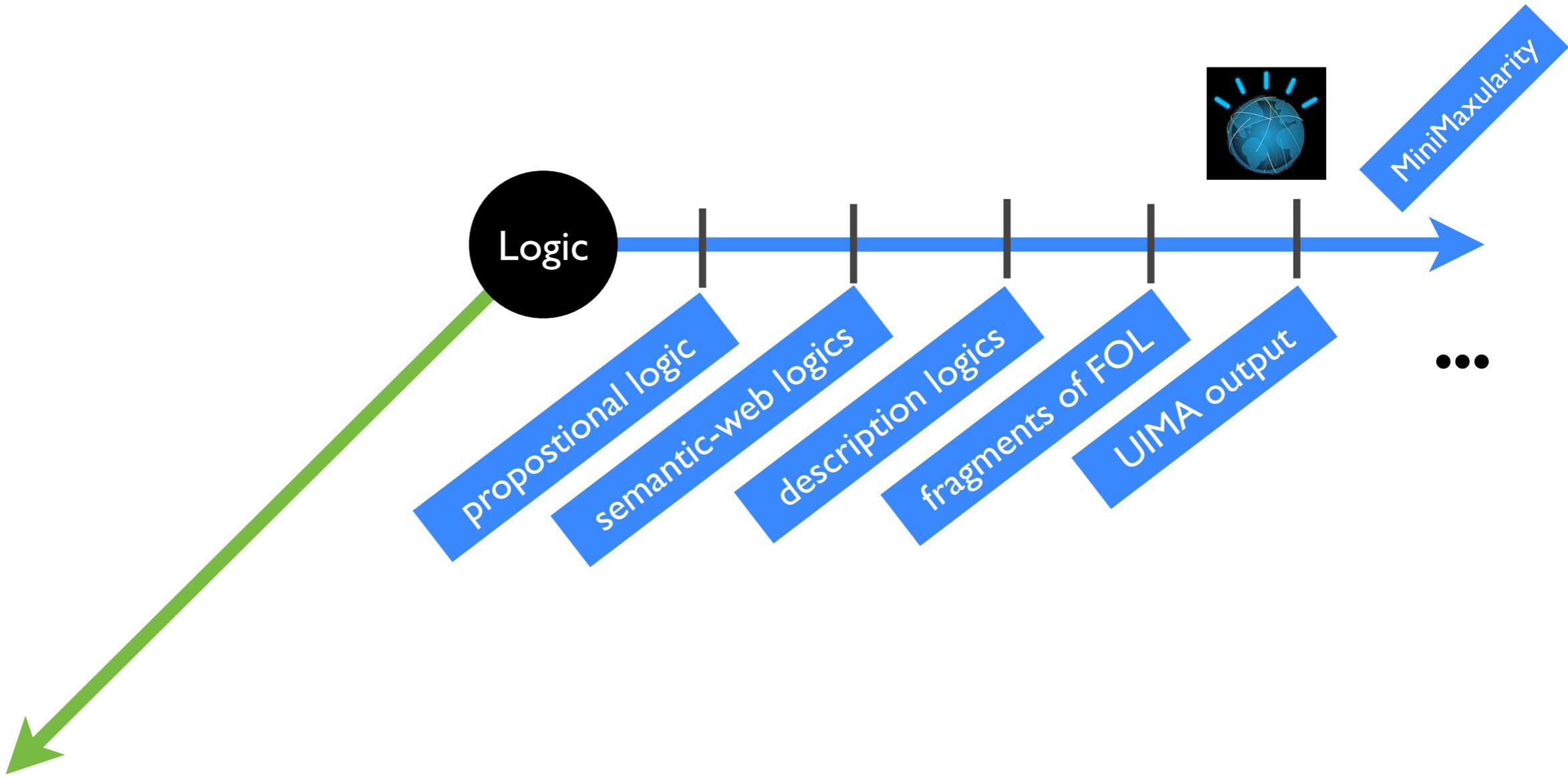






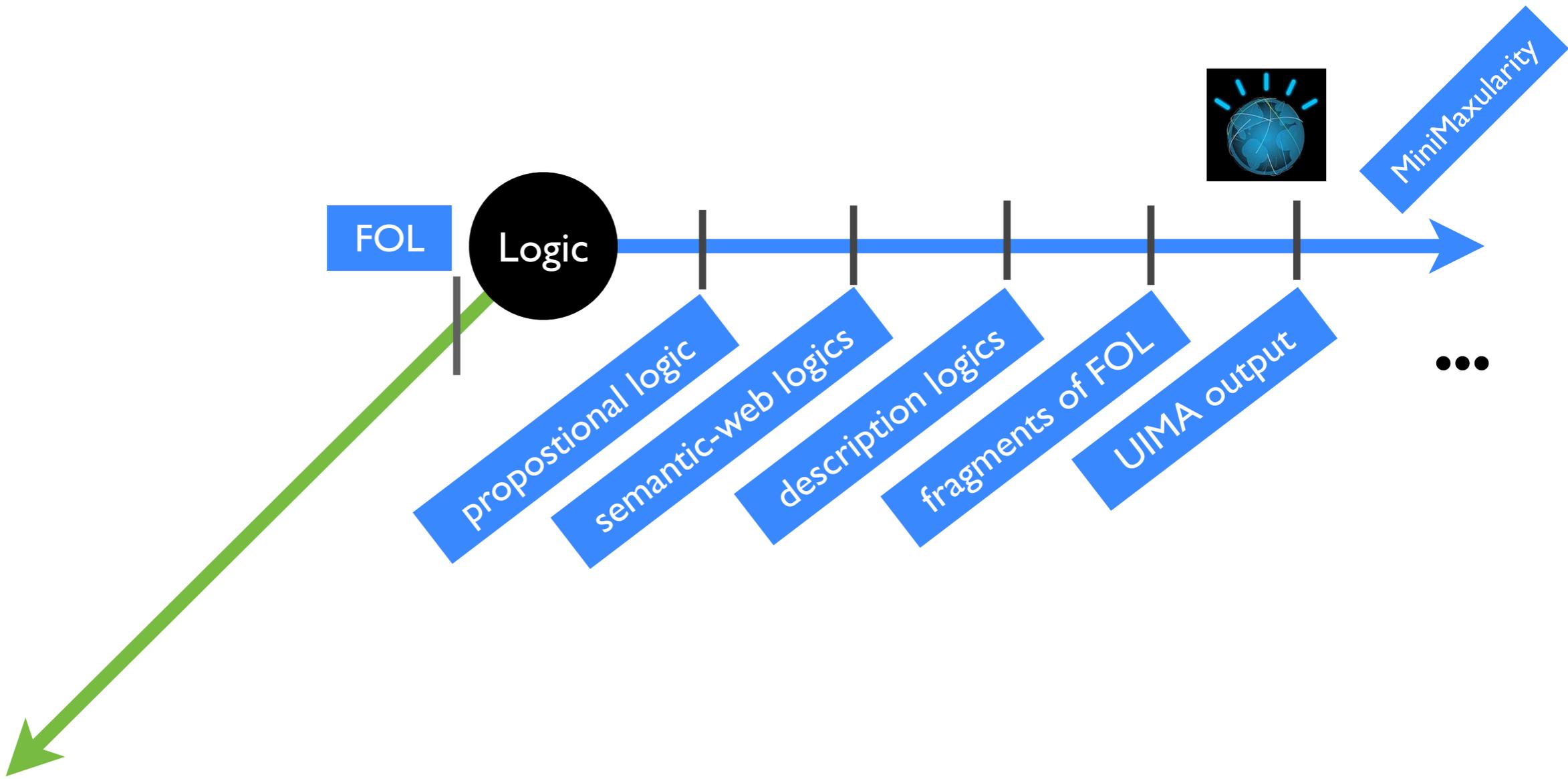


Art of Infallibility I



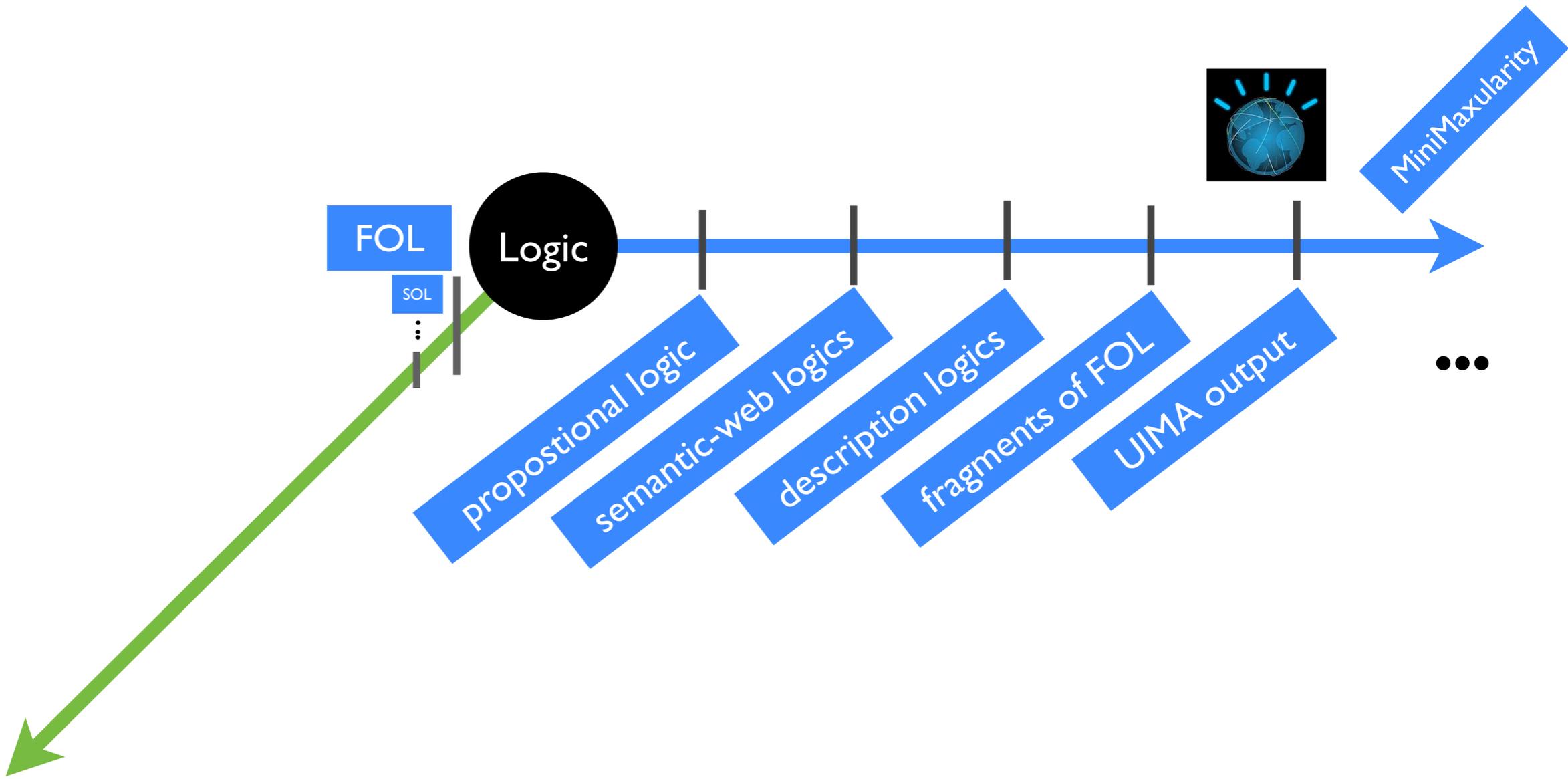


Art of Infallibility I



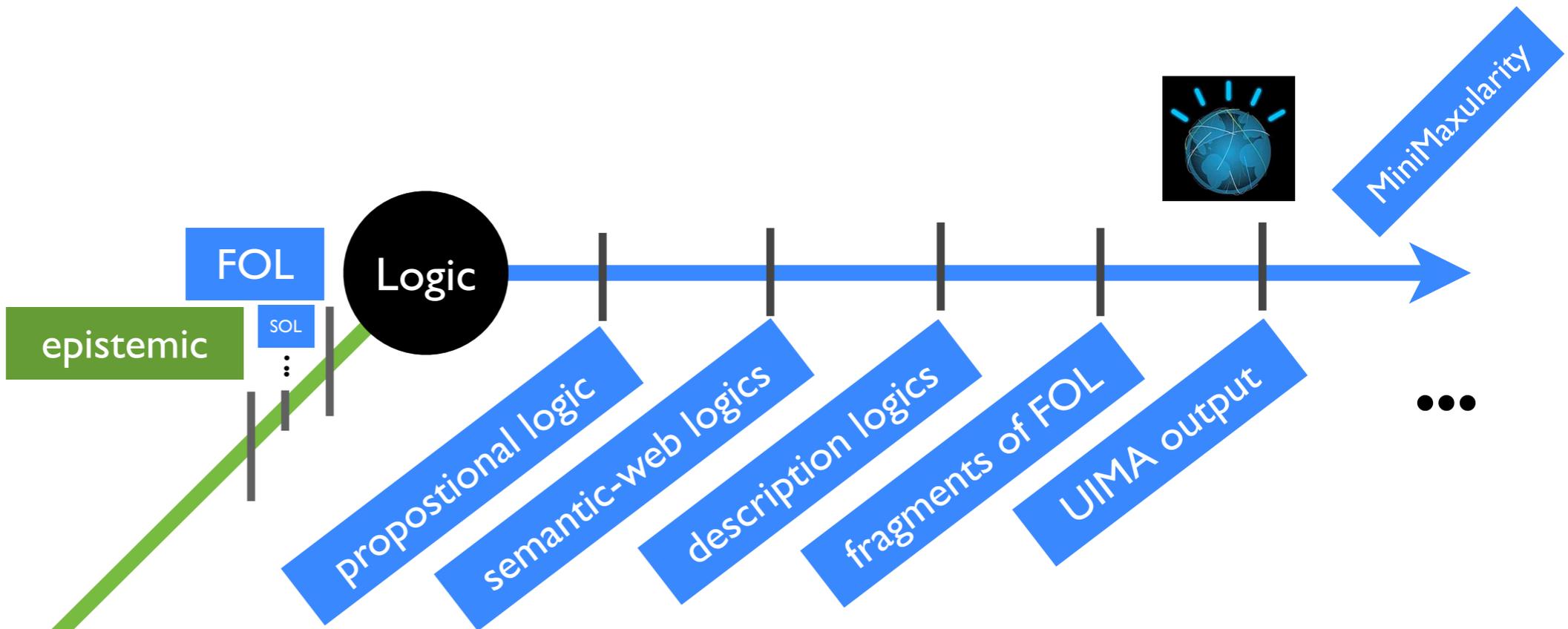


Art of Infallibility I



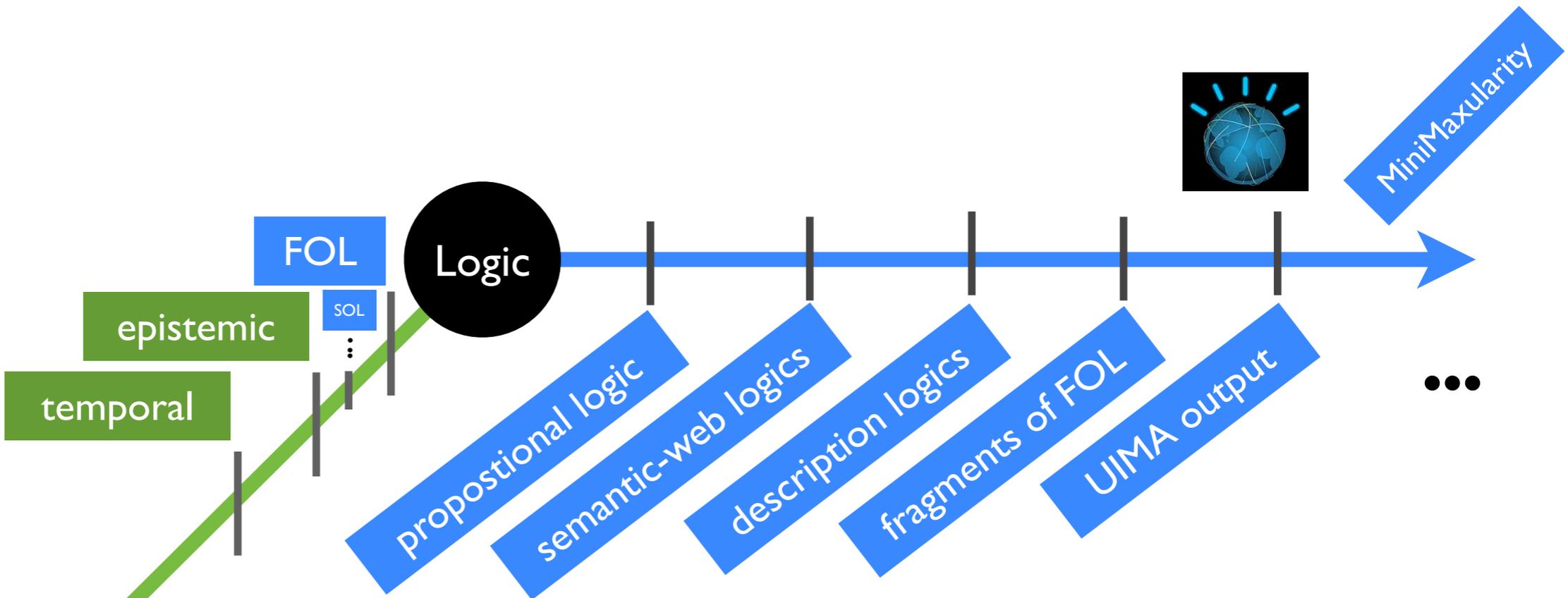


Art of Infallibility I





Art of Infallibility I





Art of Infallibility I

temporal+epistemic

temporal

epistemic

FOL

SOL

⋮

Logic

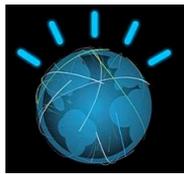
propositional logic

semantic-web logics

description logics

fragments of FOL

UIMA output



MiniMaxularity

⋮



Art of Infallibility I

temporal+epistemic+deontic

temporal+epistemic

temporal

epistemic

FOL

SOL

Logic

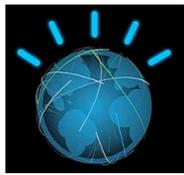
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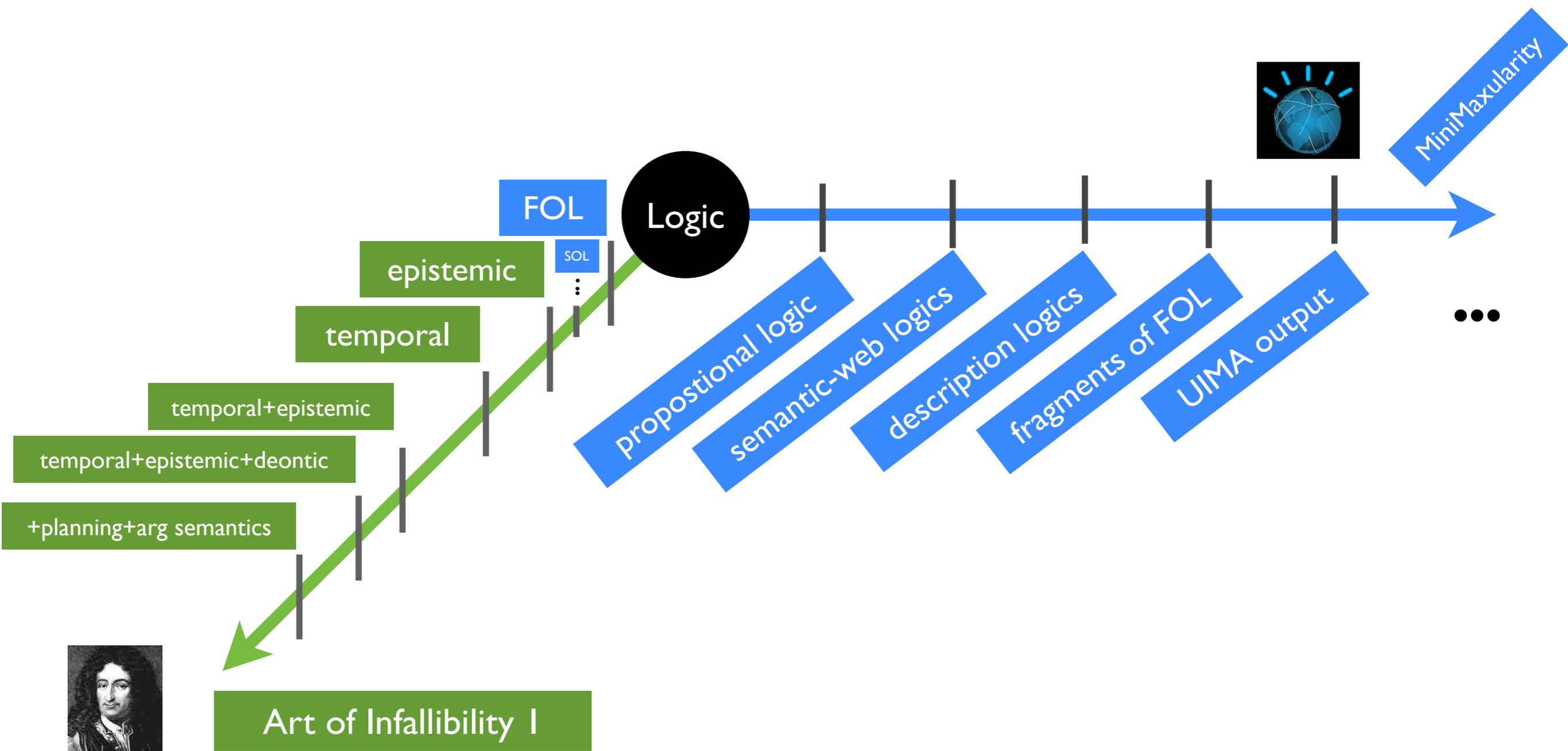
fragments of FOL

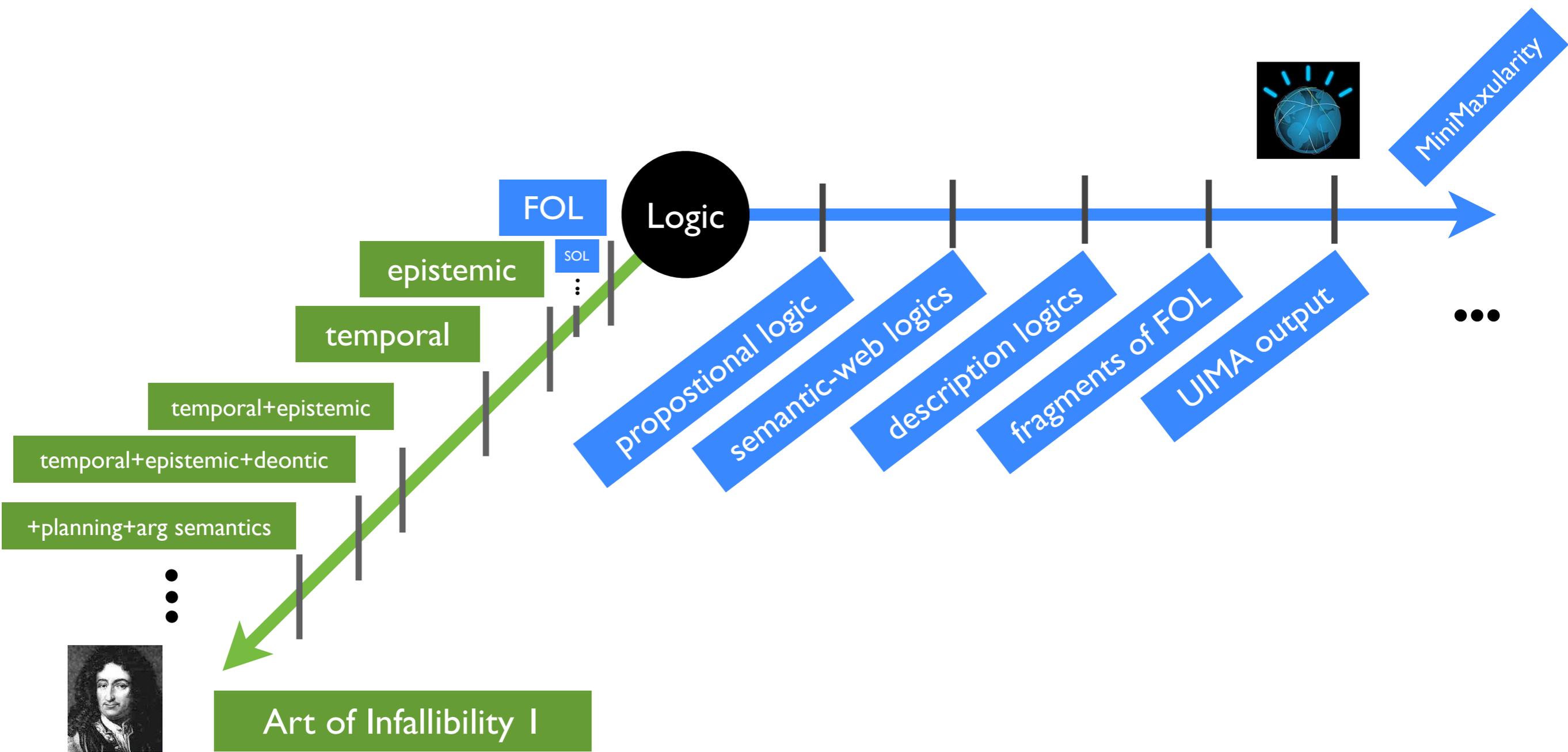
UIMA output

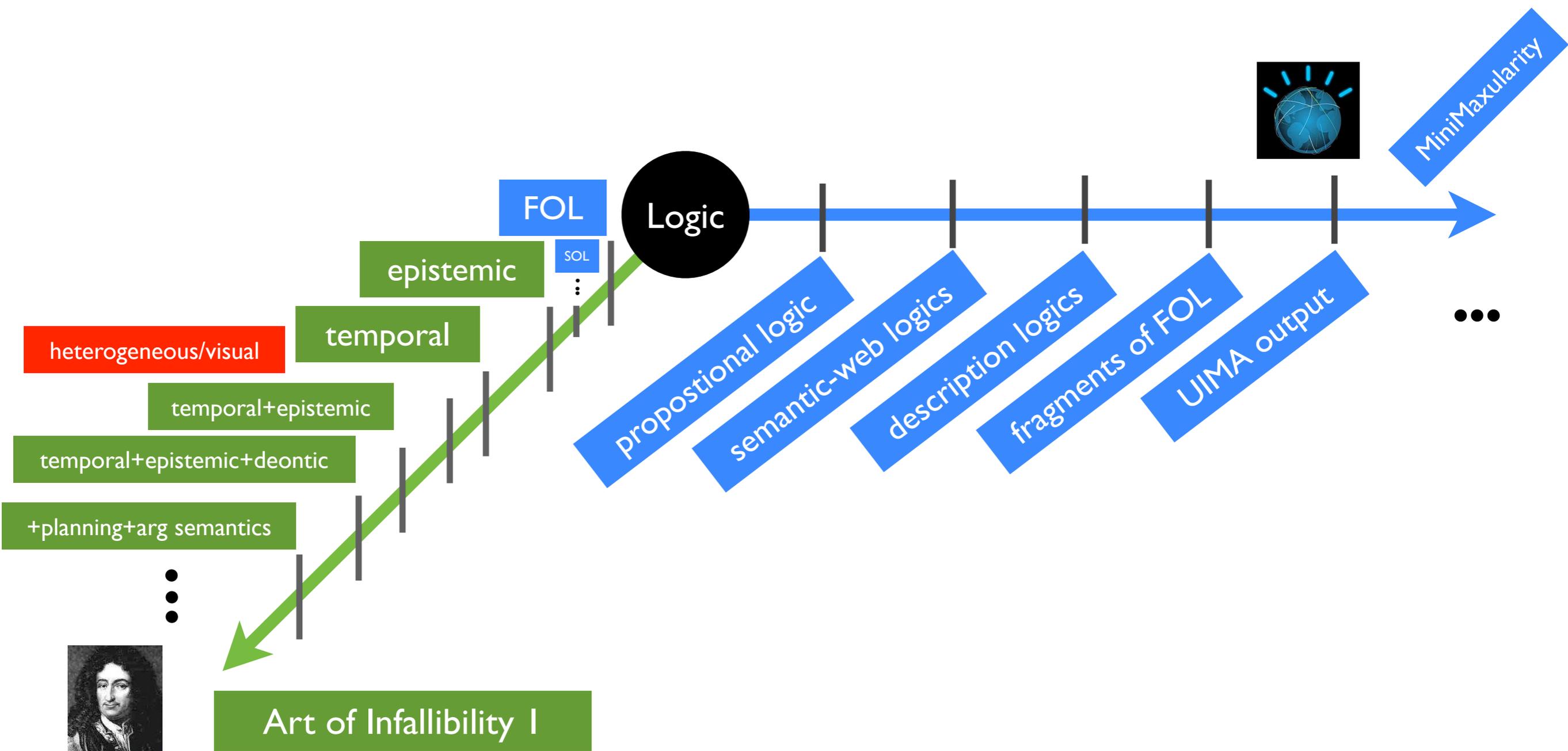


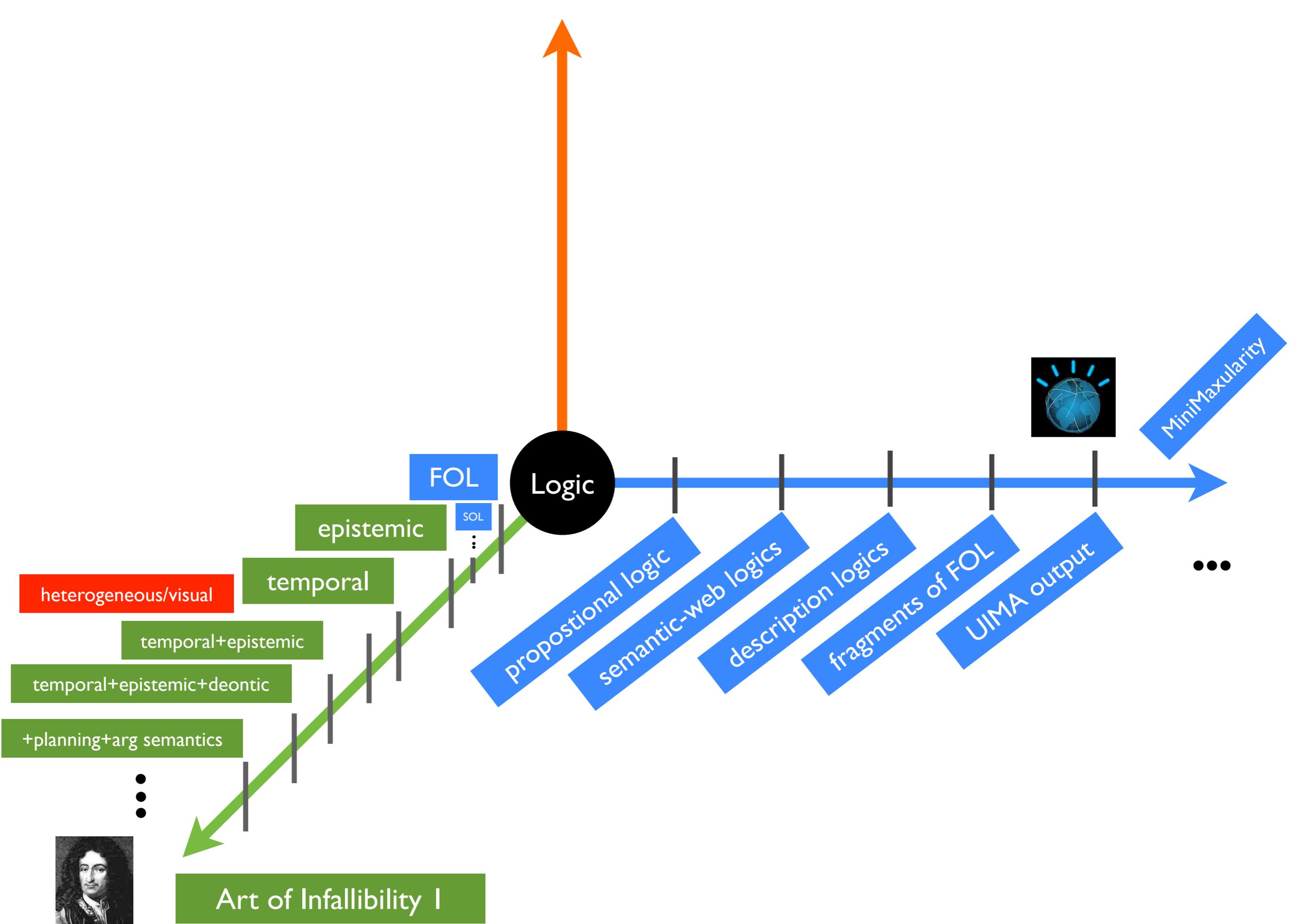
MiniMaxularity

...









Infinitary (AoI 2)



Logic

FOL

SOL

epistemic

temporal

heterogeneous/visual

temporal+epistemic

temporal+epistemic+deontic

+planning+arg semantics

...

Art of Infallibility I



propositional logic

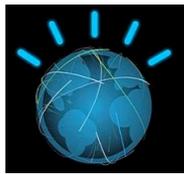
semantic-web logics

description logics

fragments of FOL

UIMA output

...



MiniMaxularity

Infinitary (AoI 2)



$L_{\omega 1, \omega}$

Logic

FOL

SOL

epistemic

temporal

heterogeneous/visual

temporal+epistemic

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⋮



Art of Infallibility I

propositional logic

semantic-web logics

description logics

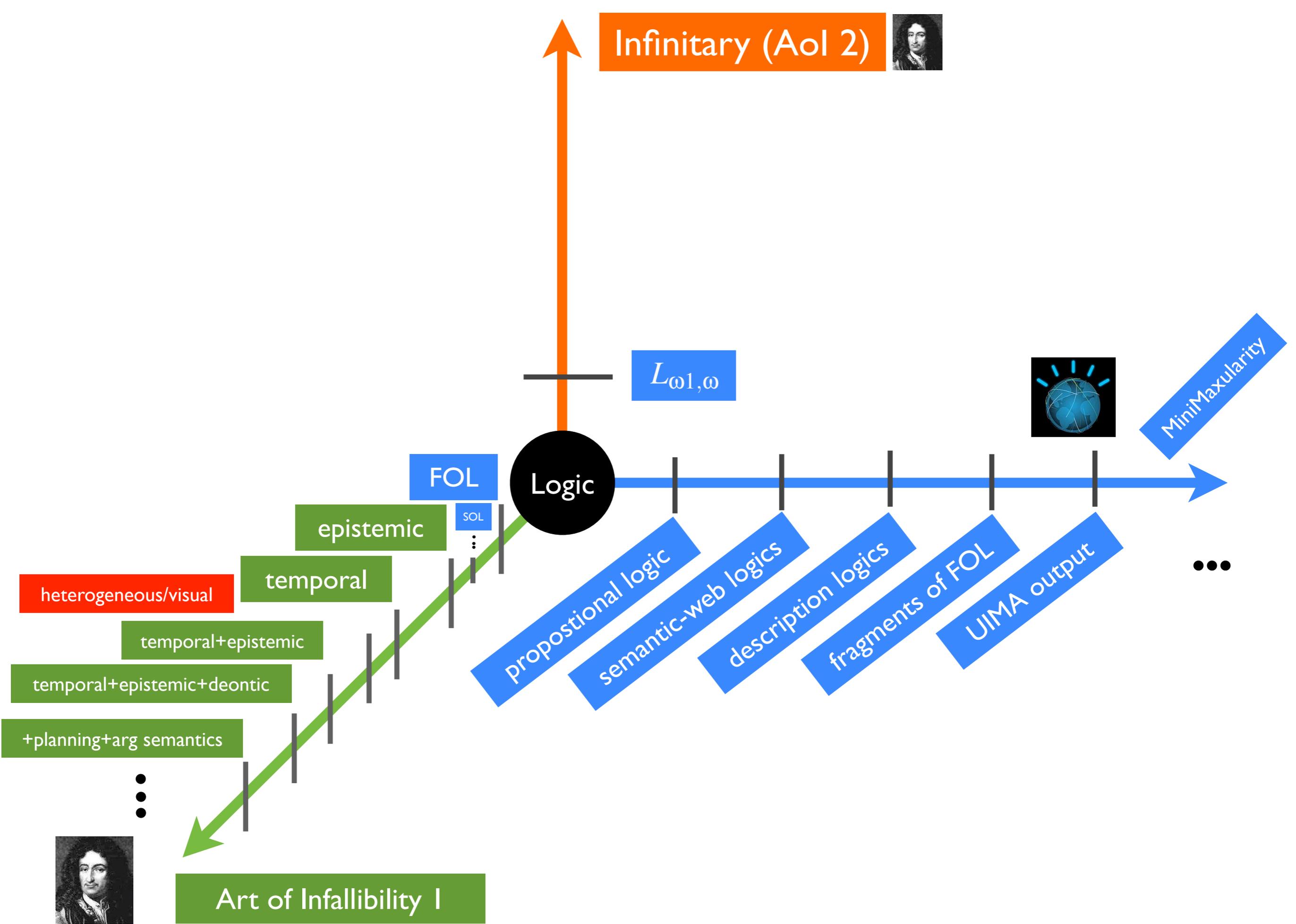
fragments of FOL

UIMA output

⋮

MiniMaxularity





Infinitary (AoI 2) 

$L_{\omega 1, \omega}$

Logic

FOL

SOL

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Art of Infallibility I

propositional logic

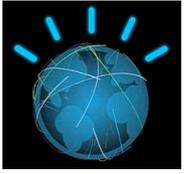
semantic-web logics

description logics

fragments of FOL

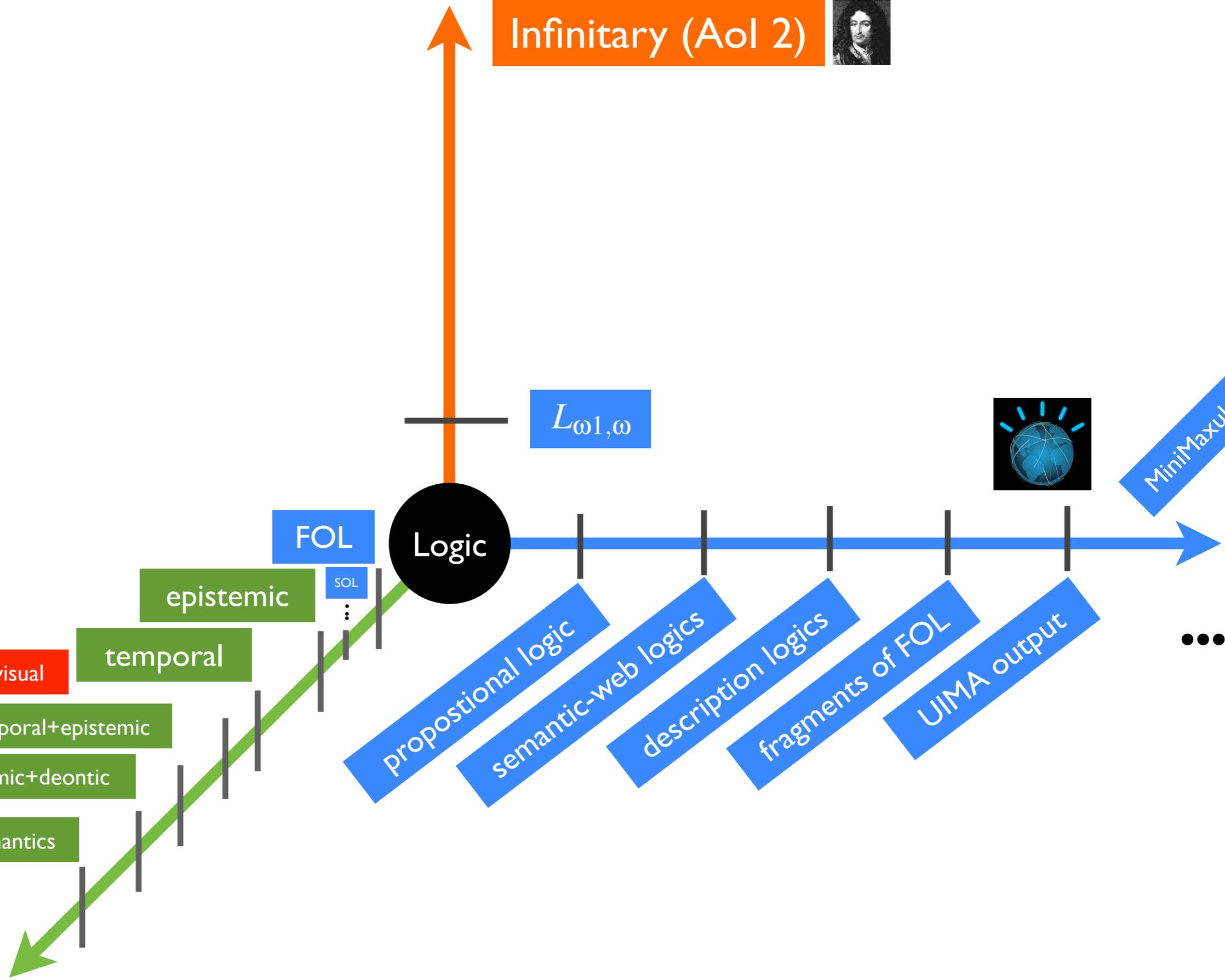
UIMA output

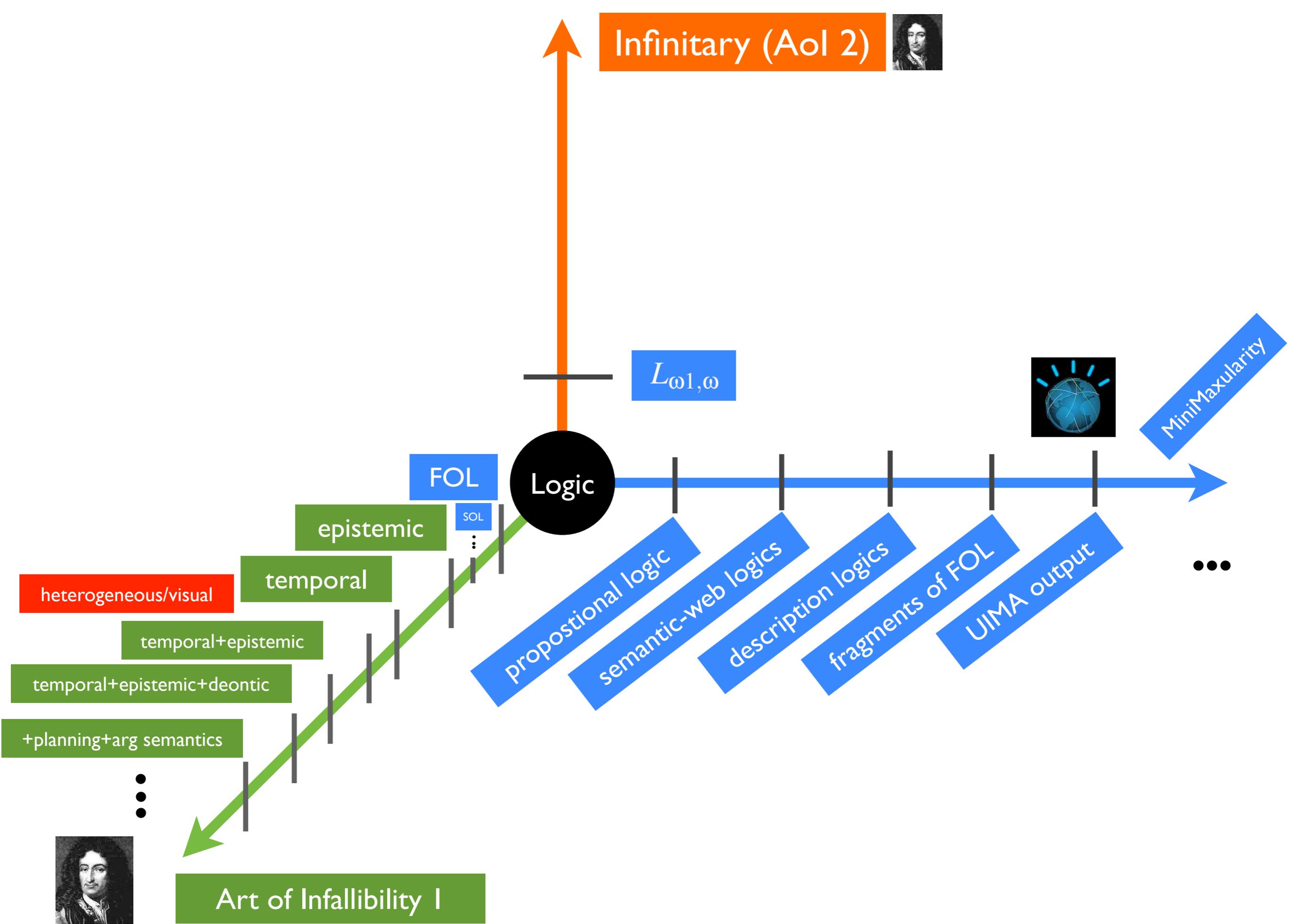
...



MiniMaxularity

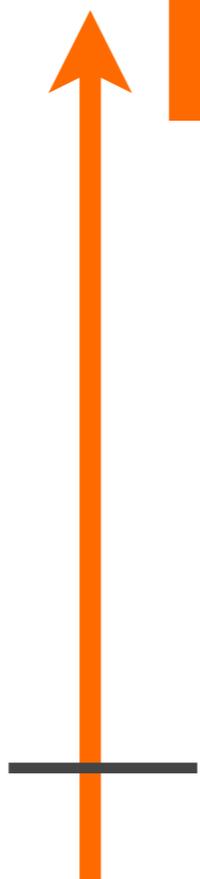
Vivid





Infinitary (AoI 2) 

$DCEC^*$   
Deontic Cognitive Event Calculus  
(with Castañeda's \*)



$L_{\omega 1, \omega}$



FOL

SOL

epistemic

temporal

heterogeneous/visual

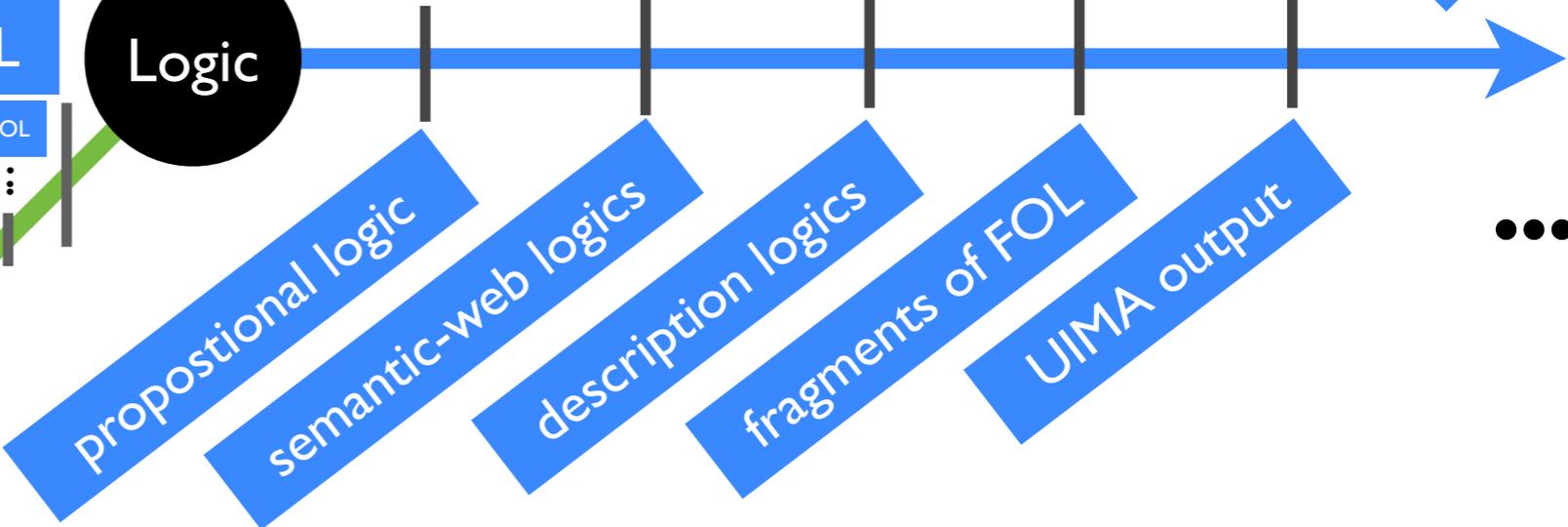
temporal+epistemic

temporal+epistemic+deontic

+planning+arg semantics



Art of Infallibility I



MiniMaxularity

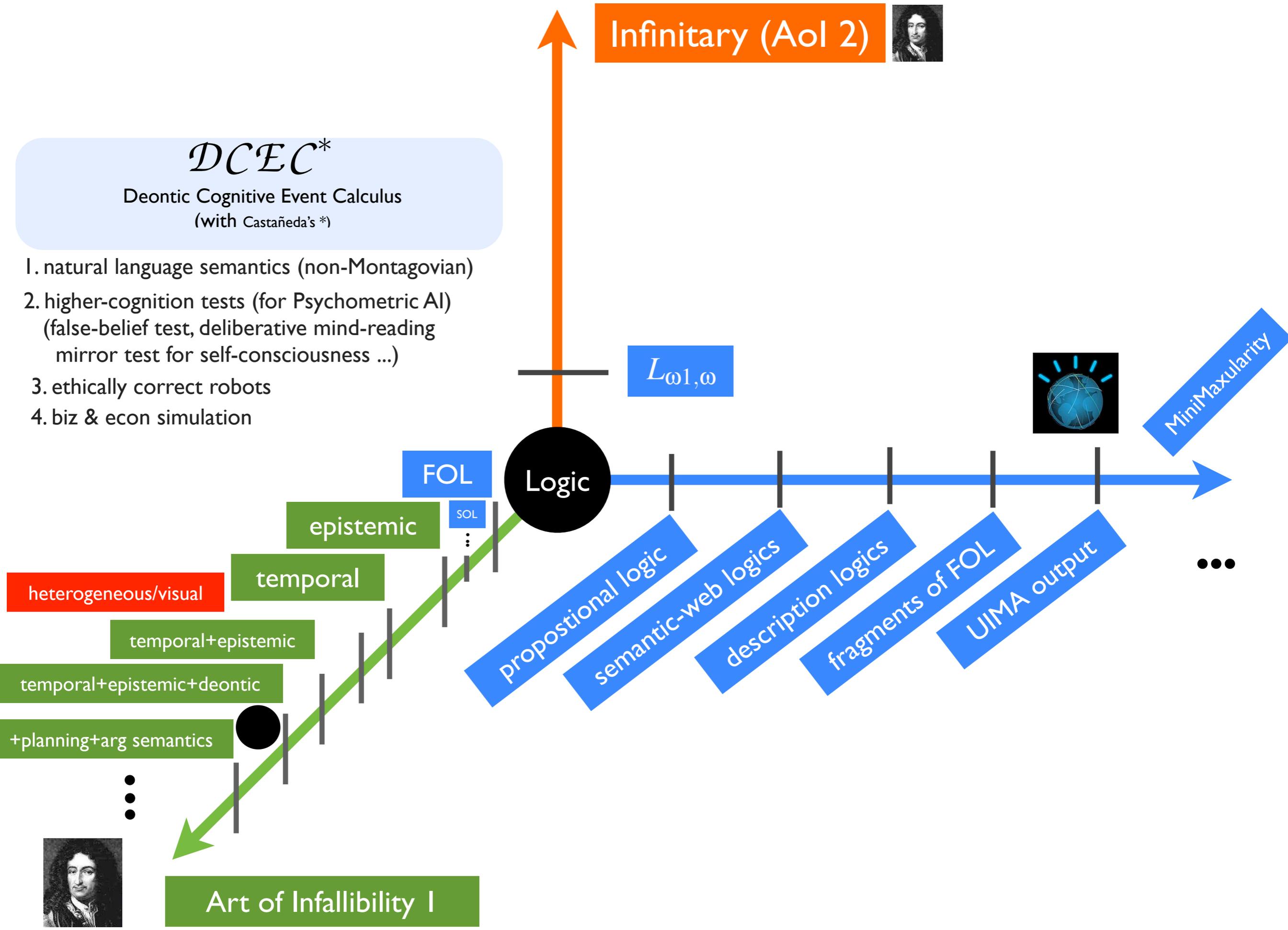
...

# Infinitary (Aol 2)

## $DCEC^*$

Deontic Cognitive Event Calculus  
(with Castañeda's \*)

- 1. natural language semantics (non-Montagovian)
- 2. higher-cognition tests (for Psychometric AI)  
(false-belief test, deliberative mind-reading mirror test for self-consciousness ...)
- 3. ethically correct robots
- 4. biz & econ simulation



Infinitary (AoI 2) 

$DCEC^*$   
Deontic Cognitive Event Calculus  
(with Castañeda's \*)



$L_{\omega 1, \omega}$



FOL

SOL

epistemic

temporal

heterogeneous/visual

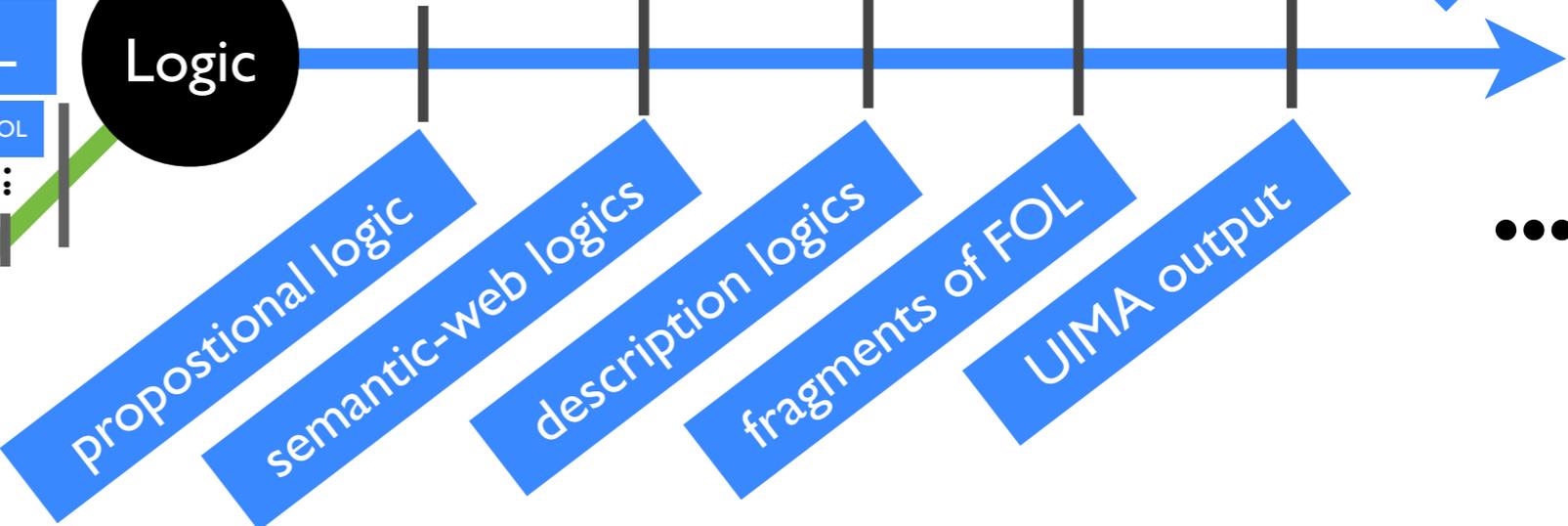
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temporal+epistemic+deontic

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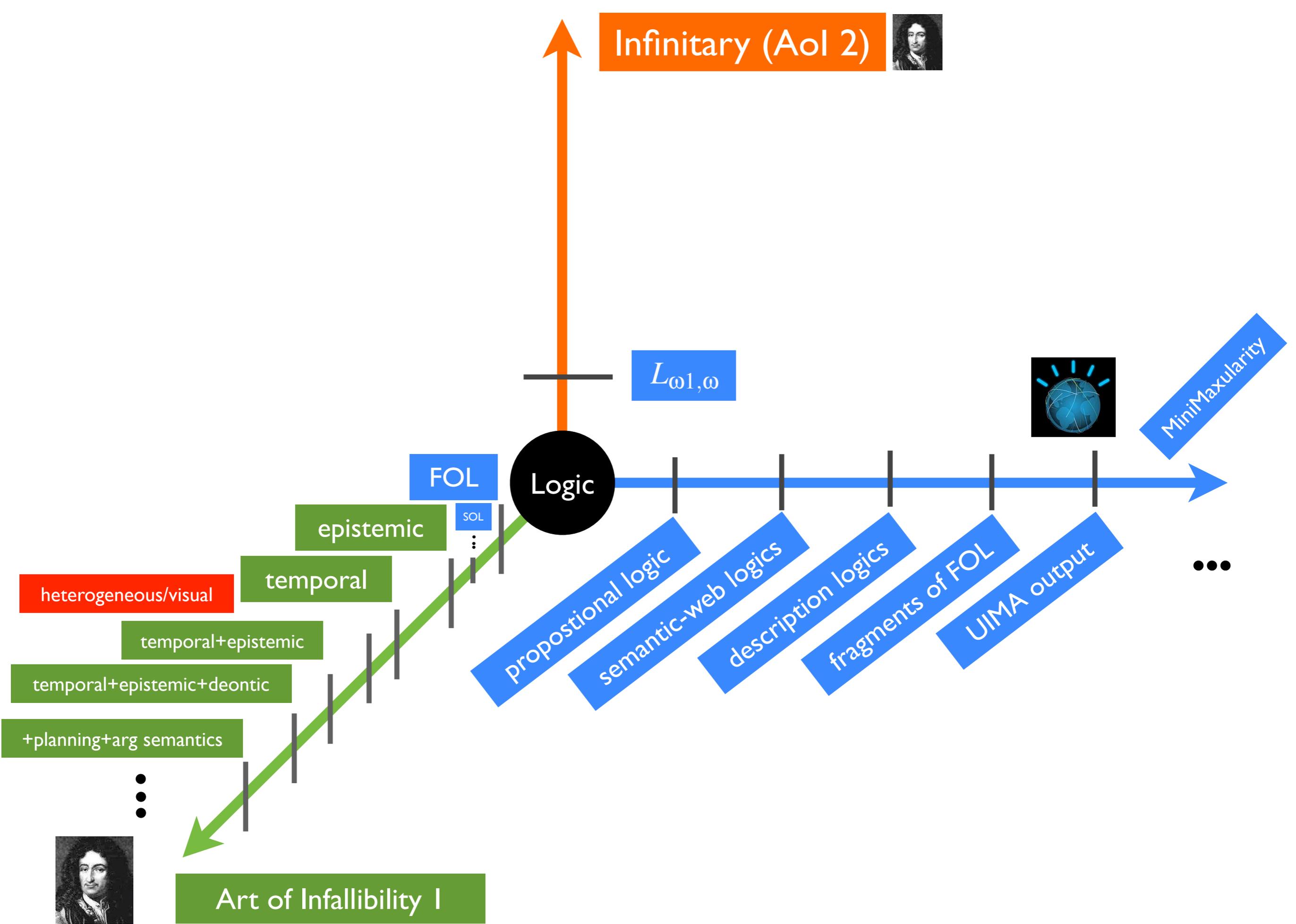


Art of Infallibility I



MiniMaxularity





Infinitary (AoI 2)



$L_{\omega 1, \omega}$

Logic

FOL

SOL

epistemic

temporal

heterogeneous/visual

temporal+epistemic

temporal+epistemic+deontic

+planning+arg semantics

⋮



Art of Infallibility I

Gödel's "God Theorem"

propositional logic

semantic-web logics

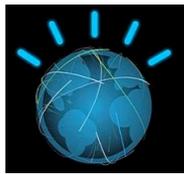
description logics

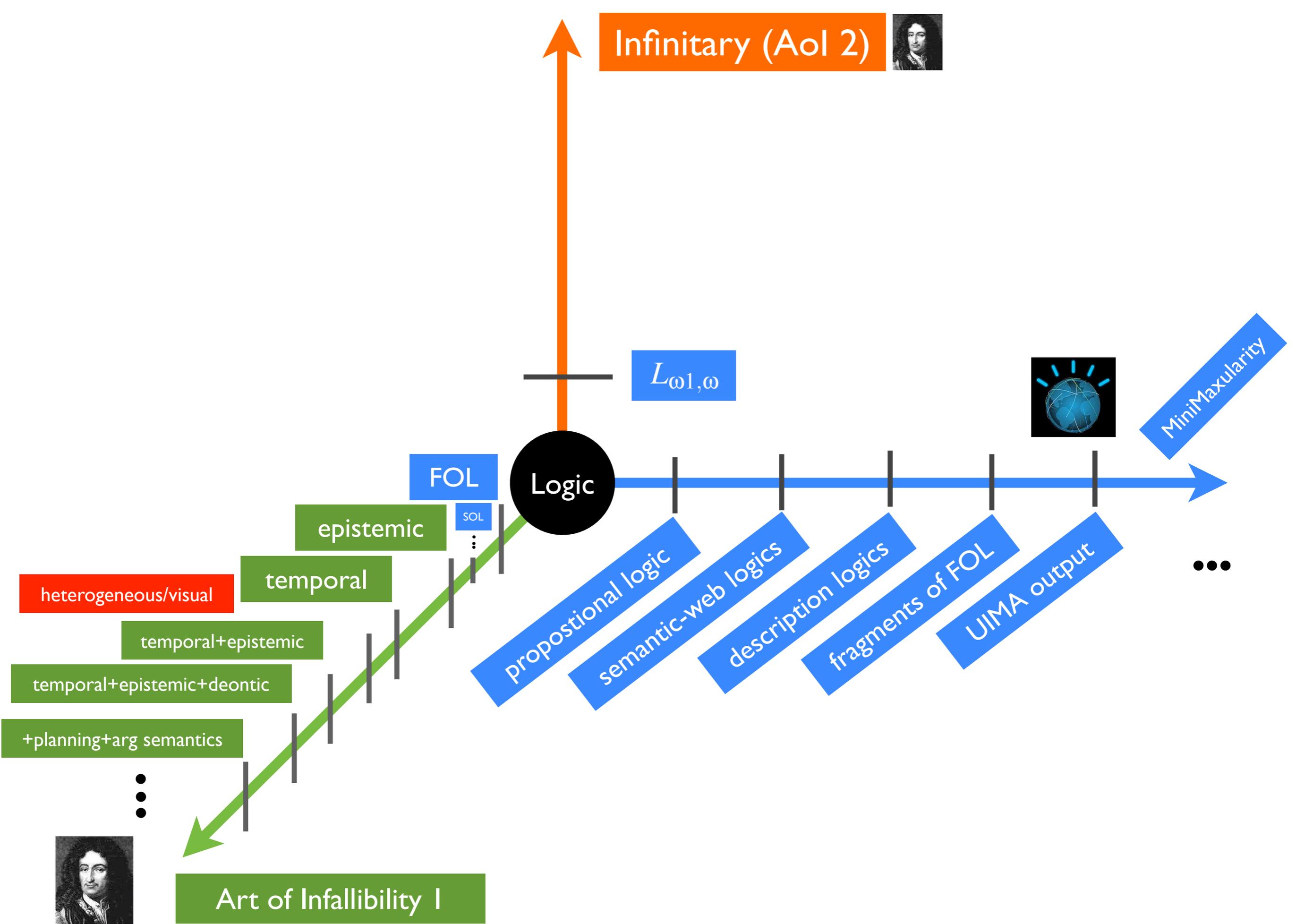
fragments of FOL

UIMA output

⋮

MiniMaxularity





Infinitary (AoI 2) 

$L_{\omega 1, \omega}$

Logic

FOL

SOL

epistemic

temporal

heterogeneous/visual

temporal+epistemic

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+planning+arg semantics

Art of Infallibility I 

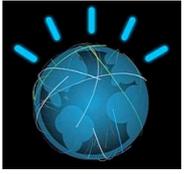
propositional logic

semantic-web logics

description logics

fragments of FOL

UIMA output

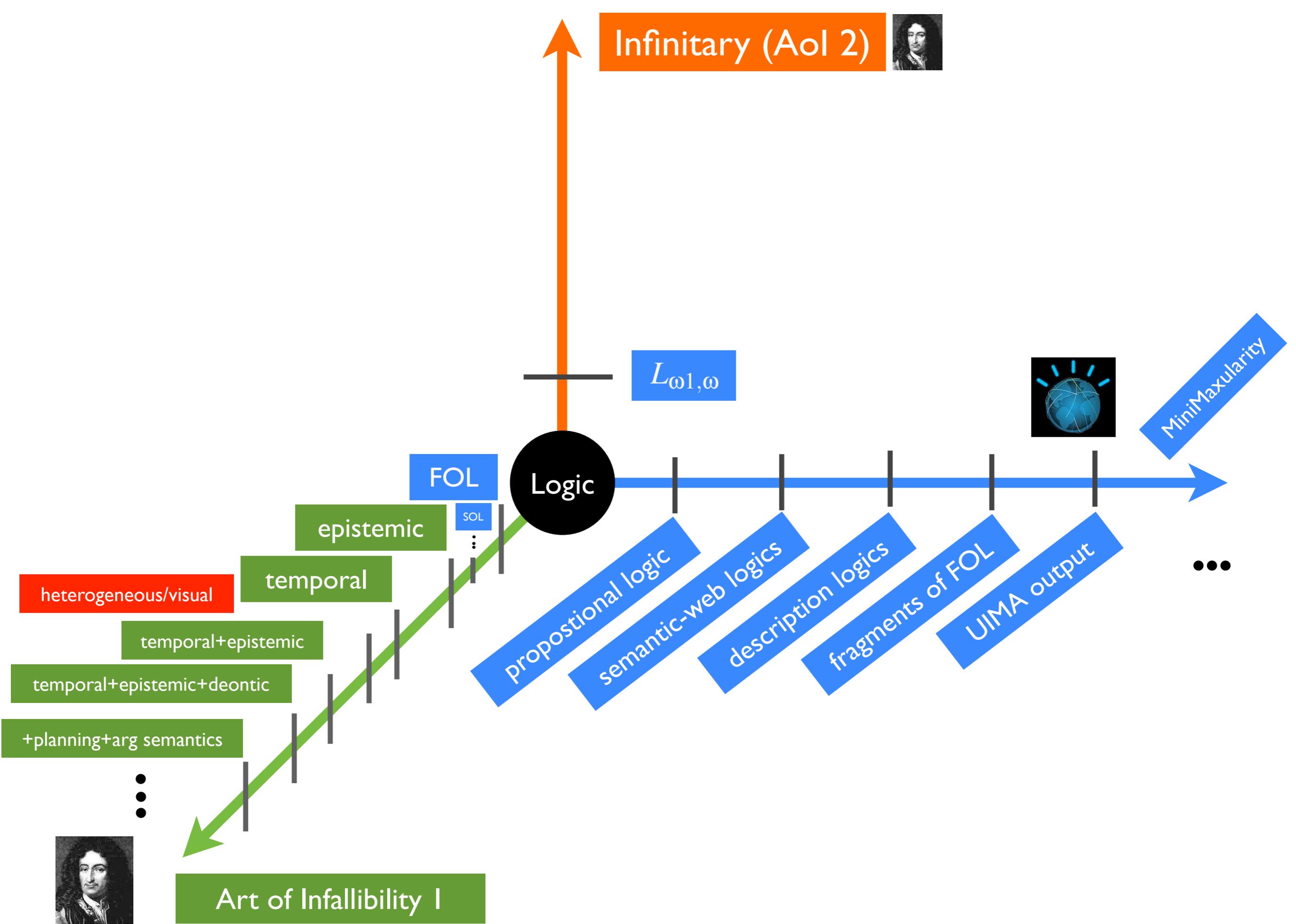


MiniMaxularity

ITS (Culture, Language, Math)

...

...



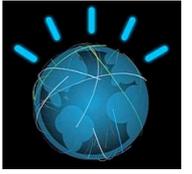
Infinitary (AoI 2) 

$L_{\omega 1, \omega}$

Logic

FOL

SOL

MiniMaxularity 

AI-ified Axiomatic Physics!  
(Synthese)

epistemic

temporal

heterogeneous/visual

temporal+epistemic

temporal+epistemic+deontic

+planning+arg semantics

Art of Infallibility I 

propositional logic

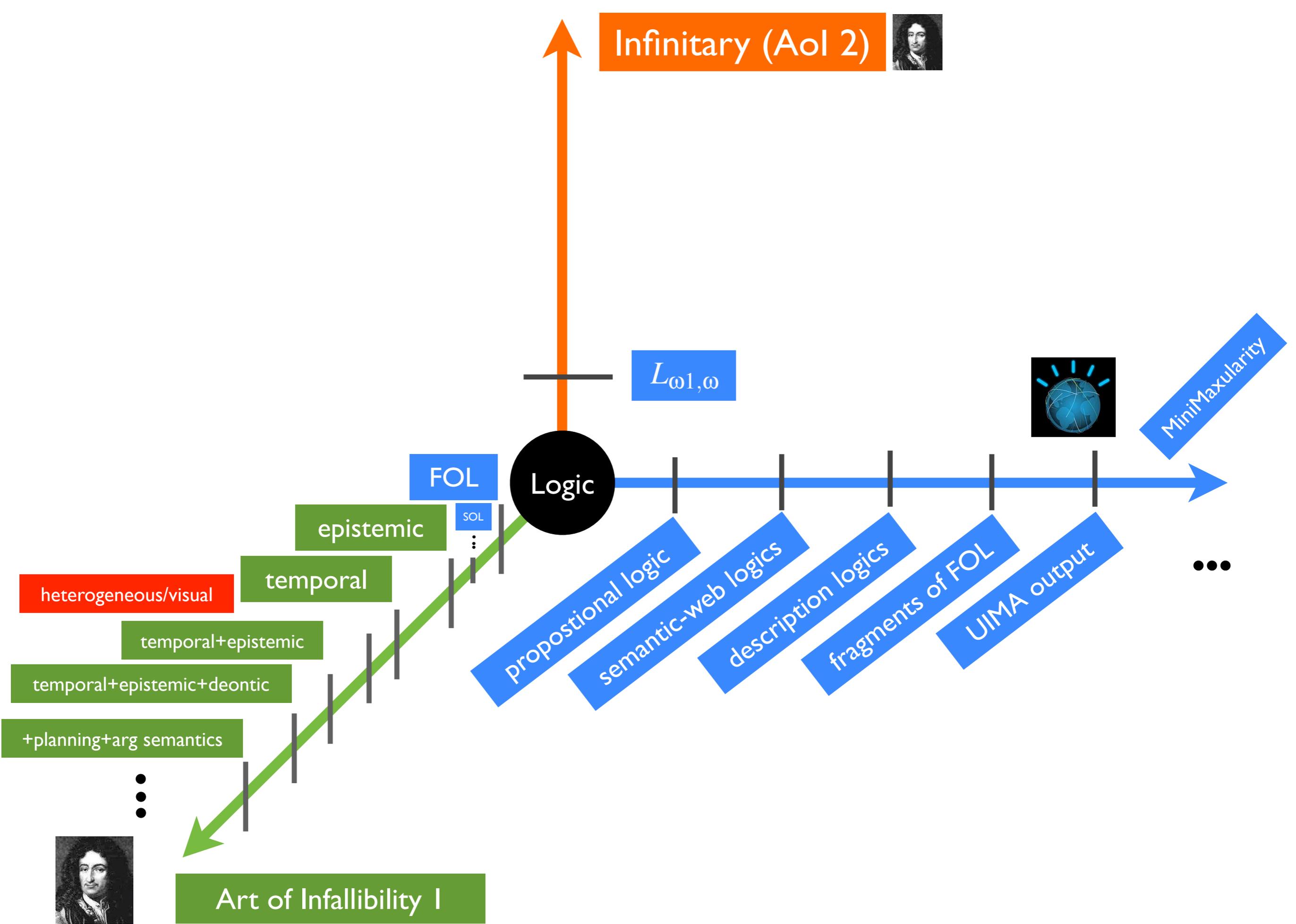
semantic-web logics

description logics

fragments of FOL

UIMA output

...



Infinitary (AoI 2)



Goodstein's Theorem!



$L_{\omega_1, \omega}$

Logic

FOL

SOL

epistemic

temporal

heterogeneous/visual

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temporal+epistemic+deontic

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Art of Infallibility I

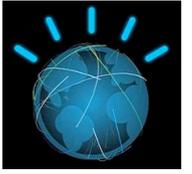
propositional logic

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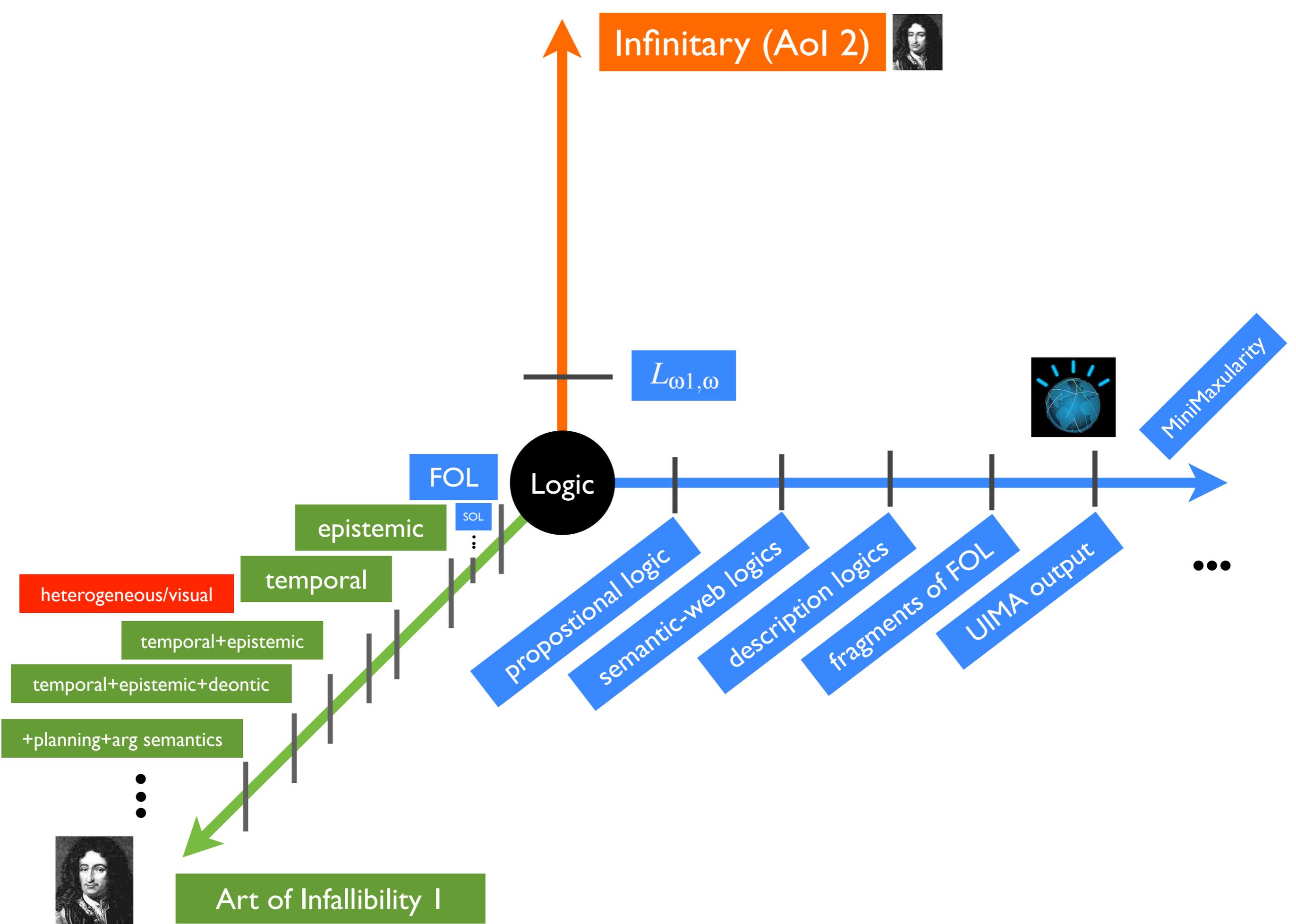
fragments of FOL

UIMA output



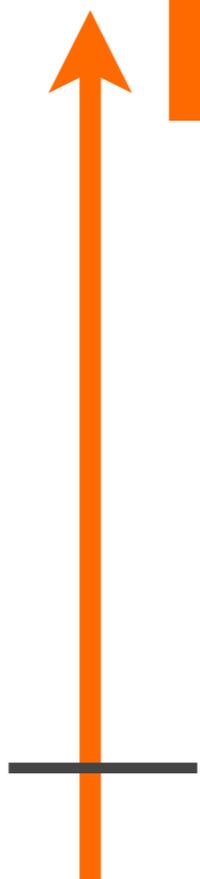
MiniMaxularity





Infinitary (AoI 2) 

$DCEC^*$   
Deontic Cognitive Event Calculus  
(with Castañeda's \*)



$L_{\omega 1, \omega}$



FOL

SOL

⋮

epistemic

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Art of Infallibility I

propositional logic

semantic-web logics

description logics

fragments of FOL

UIMA output

⋮



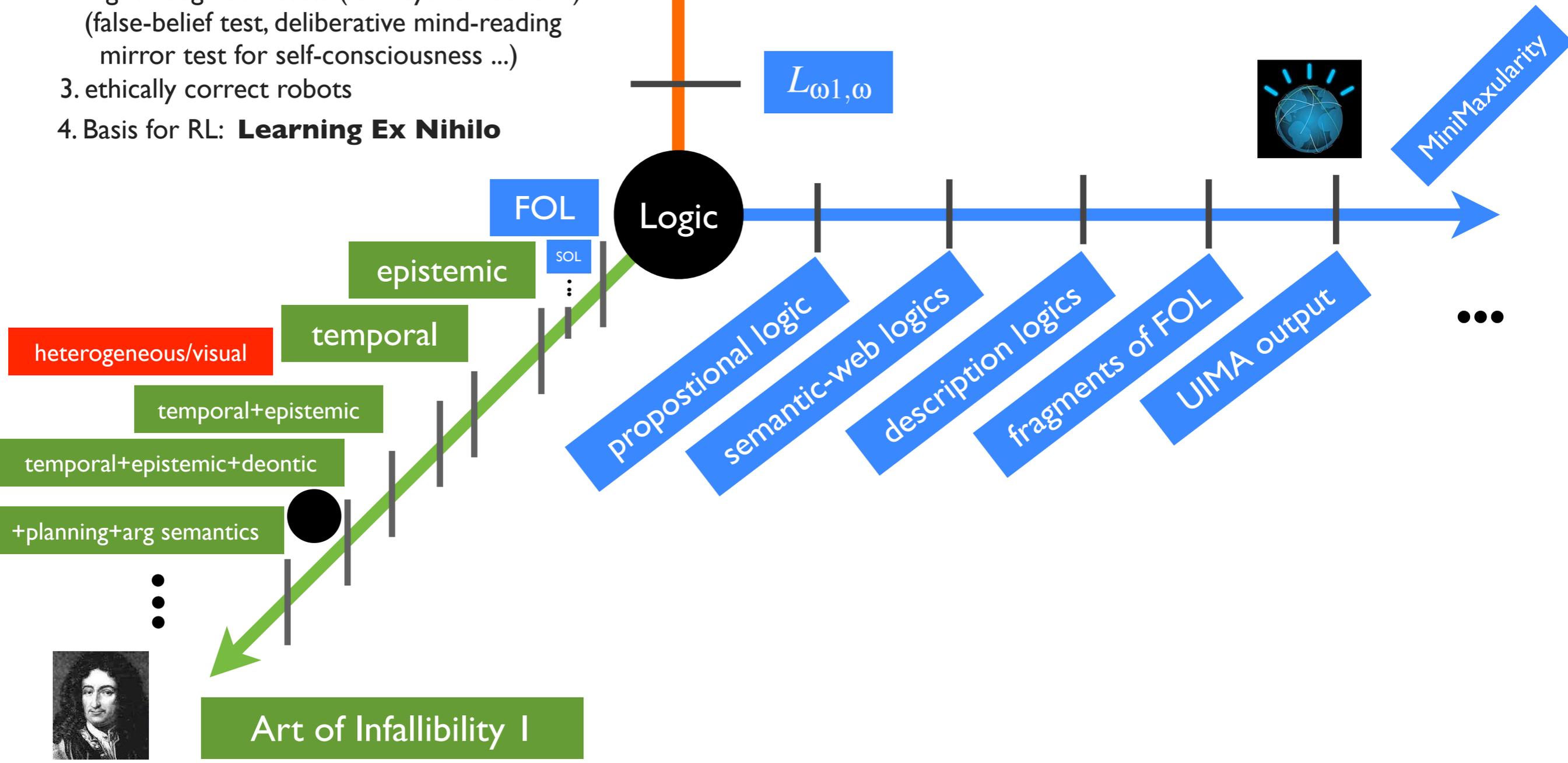
MiniMaxularity

# Infinitary (Aol 2)

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Deontic Cognitive Event Calculus  
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- 3. ethically correct robots
- 4. Basis for RL: **Learning Ex Nihilo**

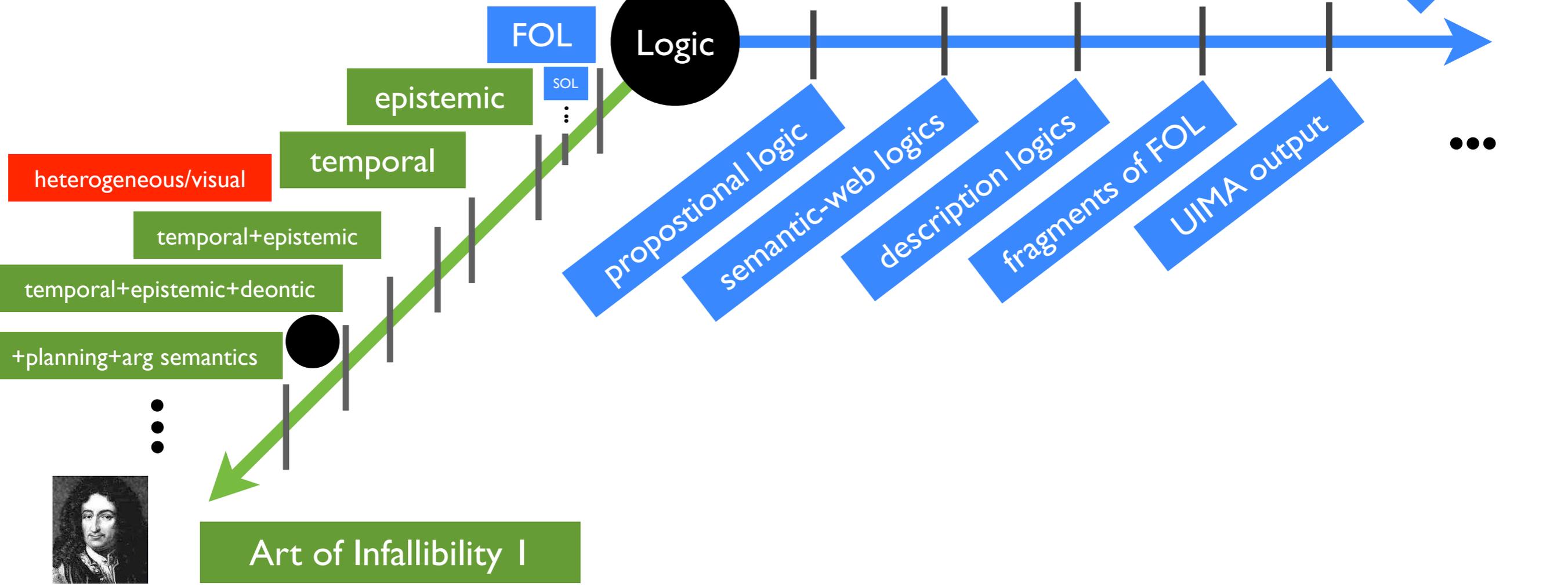


# Infinitary (Aol 2)

## $DCEC^*$

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4. Basis for RL: **Learning Ex Nihilo**



# Animal-Level AI

*Super-Serious* Human Cognitive Power

*Serious* Human Cognitive Power

Mere Calculative Cognitive Power

**Entscheidungsproblem**

# Animal-Level AI

Analytical Hierarchy

*Serious Human Cognitive Power*

Mere Calculative Cognitive Power

**Entscheidungsproblem**

# Animal-Level AI

Analytical Hierarchy

Arithmetical Hierarchy

**Entscheidungsproblem**

Mere Calculative Cognitive Power

# Animal-Level AI

Analytical Hierarchy

Arithmetical Hierarchy

Polynomial Hierarchy

**Entscheidungsproblem**

# Animal-Level AI

Analytical Hierarchy

Arithmetical Hierarchy

**Entscheidungsproblem**

Polynomial Hierarchy

$P \subseteq NP \subseteq PSPACE = NPSpace \subseteq EXPTIME \subseteq NEXPTIME \subseteq EXPSPACE$

# Animal-Level AI

## Analytical Hierarchy

### Arithmetical Hierarchy

$\vdots$   
 $\Pi_2$   
 $\Sigma_2$   
 $\Pi_1$   
 $\Sigma_1$   
 $\Sigma_0$

**Entscheidungsproblem**

### Polynomial Hierarchy

$P \subseteq NP \subseteq PSPACE = NPSpace \subseteq EXPTIME \subseteq NEXPTIME \subseteq EXPSPACE$

# Animal-Level AI

## Analytical Hierarchy

### Arithmetical Hierarchy

Go:AlphaGo



⋮  
 $\Pi_2$   
 $\Sigma_2$   
 $\Pi_1$   
 $\Sigma_1$   
 $\Sigma_0$

**Entscheidungsproblem**

### Polynomial Hierarchy

$P \subseteq NP \subseteq PSPACE = NPSpace \subseteq EXPTIME \subseteq NEXPTIME \subseteq EXPSPACE$

# Animal-Level AI

## Analytical Hierarchy

### Arithmetical Hierarchy

*Jeopardy!:* **Watson**

*Go:* AlphaGo



⋮  
 $\Pi_2$   
 $\Sigma_2$   
 $\Pi_1$   
 $\Sigma_1$   
 $\Sigma_0$

**Entscheidungsproblem**

### Polynomial Hierarchy

$P \subseteq NP \subseteq PSPACE = NPSpace \subseteq EXPTIME \subseteq NEXPTIME \subseteq EXPSPACE$

# Animal-Level AI

## Analytical Hierarchy

### Arithmetical Hierarchy

Chess: Deep Blue



Jeopardy!: **Watson**



Go: AlphaGo



⋮  
 $\Pi_2$   
 $\Sigma_2$   
 $\Pi_1$   
 $\Sigma_1$   
 $\Sigma_0$

**Entscheidungsproblem**

### Polynomial Hierarchy

$P \subseteq NP \subseteq PSPACE = NPSpace \subseteq EXPTIME \subseteq NEXPTIME \subseteq EXPSPACE$

# Animal-Level AI

## Analytical Hierarchy

### Arithmetical Hierarchy

Checkers: Chinook



Chess: Deep Blue



Jeopardy!: **Watson**



Go: AlphaGo



⋮  
 $\Pi_2$   
 $\Sigma_2$   
 $\Pi_1$   
 $\Sigma_1$   
 $\Sigma_0$

**Entscheidungsproblem**

### Polynomial Hierarchy

$P \subseteq NP \subseteq PSPACE = NPSpace \subseteq EXPTIME \subseteq NEXPTIME \subseteq EXPSPACE$

# Animal-Level AI

## Analytical Hierarchy

### Arithmetical Hierarchy

$\vdots$   
 $\Pi_2$   
 $\Sigma_2$   
 $\Pi_1$   
 $\Sigma_1$   
 $\Sigma_0$

**Entscheidungsproblem**

### Polynomial Hierarchy

Jeopardy!: **Watson**

Chess: Deep Blue  
Checkers: Chinook  
Go: AlphaGo

$P \subseteq NP \subseteq PSPACE = NPSPACE \subseteq EXPTIME \subseteq NEXPTIME \subseteq EXPSPACE$

# Animal-Level AI

## Analytical Hierarchy

### Arithmetical Hierarchy

⋮  
 $\Pi_2$   
 $\Sigma_2$   
 $\Pi_1$   
 $\Sigma_1$   
 $\Sigma_0$

**Entscheidungsproblem**

### Polynomial Hierarchy

Jeopardy!: **Watson**

Chess: Deep Blue  
Checkers: Chinook  
Go: AlphaGo



$P \subseteq NP \subseteq PSPACE = NPSpace \subseteq EXPTIME \subseteq NEXPTIME \subseteq EXPSPACE$

# Animal-Level AI

## Analytical Hierarchy

### Arithmetical Hierarchy



Church

⋮  
 $\Pi_2$   
 $\Sigma_2$   
 $\Pi_1$   
 $\Sigma_1$   
 $\Sigma_0$

**Entscheidungsproblem**

### Polynomial Hierarchy

Jeopardy!: **Watson**



$P \subseteq NP \subseteq PSPACE = NPSPACE \subseteq EXPTIME \subseteq NEXPTIME \subseteq EXPSPACE$

# Animal-Level AI

## Analytical Hierarchy

### Arithmetical Hierarchy



Church



Turing

⋮  
 $\Pi_2$   
 $\Sigma_2$   
 $\Pi_1$   
 $\Sigma_1$   
 $\Sigma_0$

**Entscheidungsproblem**

### Polynomial Hierarchy

Jeopardy!: **Watson**

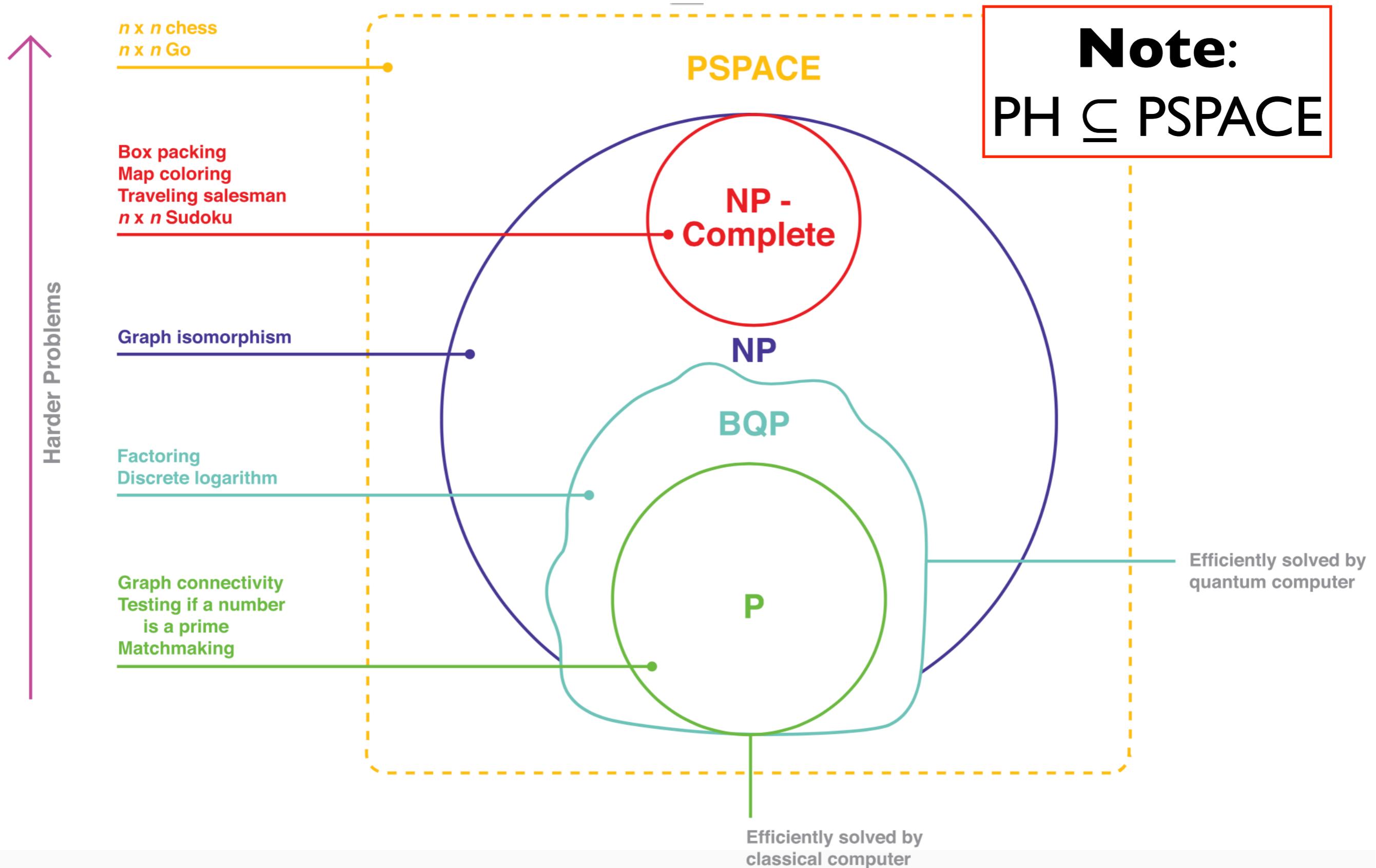


$P \subseteq NP \subseteq PSPACE = NPSPACE \subseteq EXPTIME \subseteq NEXPTIME \subseteq EXPSPACE$

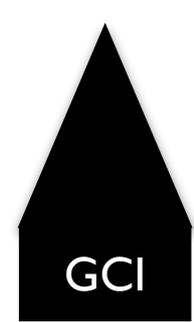
What about (oft vaunted) quantum computers?

**Note:**  
 $PH \subseteq PSPACE$

# What about (oft vaunted) quantum computers?



# What about (oft vaunted) quantum computers?



GCI

Harder Problems

*n x n* chess  
*n x n* Go

Box packing  
Map coloring  
Traveling salesman  
*n x n* Sudoku

Graph isomorphism

Factoring  
Discrete logarithm

Graph connectivity  
Testing if a number  
is a prime  
Matchmaking

PSPACE

NP -  
Complete

NP

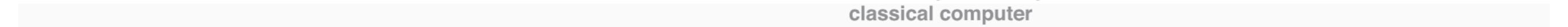
BQP

P

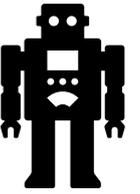
**Note:**  
 $PH \subseteq PSPACE$

Efficiently solved by  
quantum computer

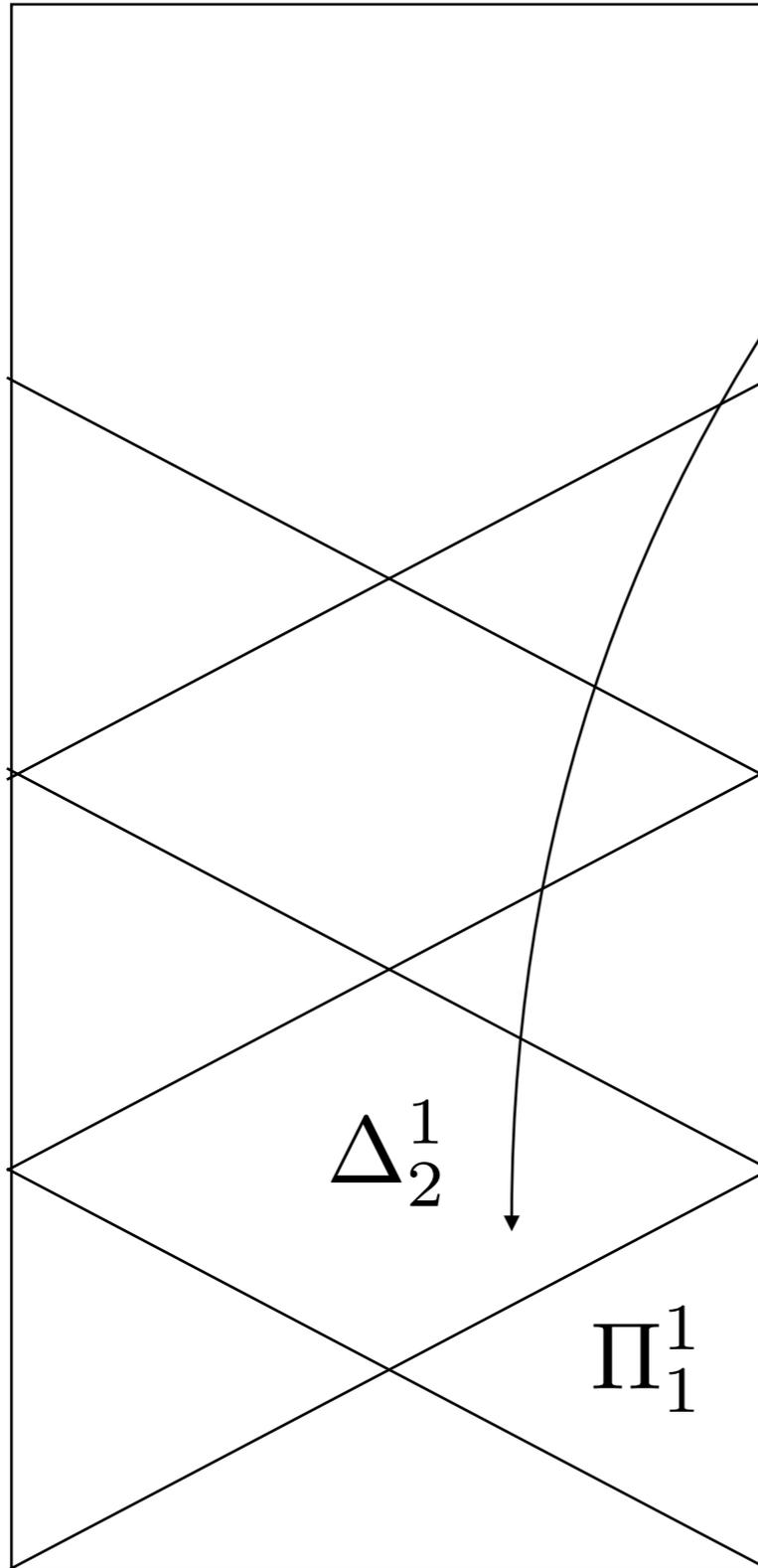
Efficiently solved by  
classical computer



CogSci and AI need to say more about where AI falls/can fall in the landscape.

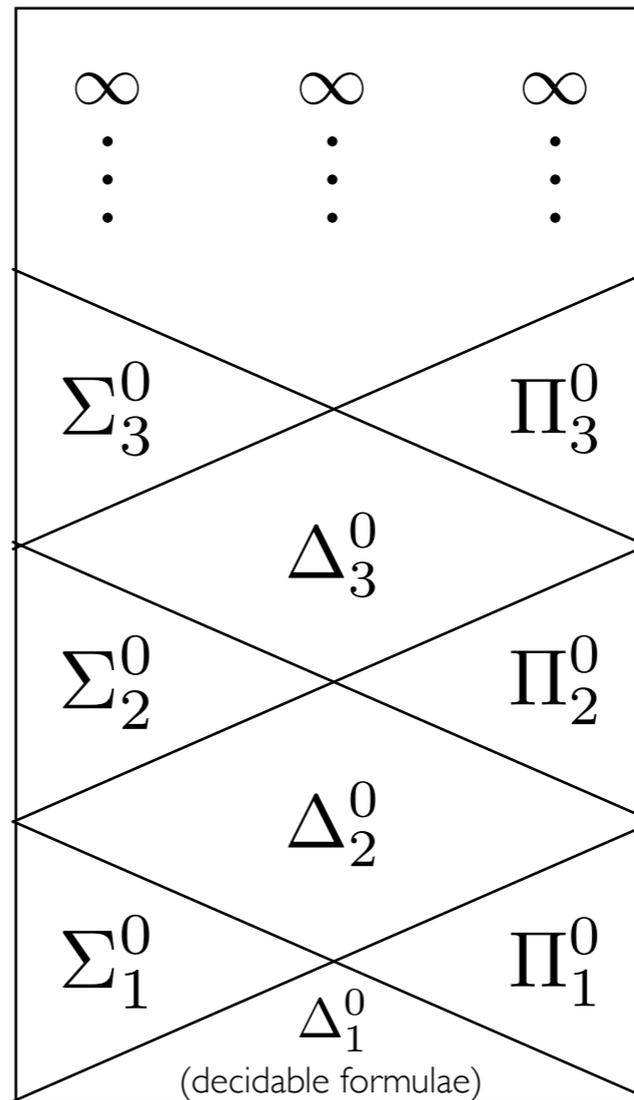


$A^n \mathcal{H}$  (Analytic Hierarchy)



Infinite Time Turing Machines (ITTMs)

$A^r \mathcal{H}$  (Arithmetic Hierarchy)

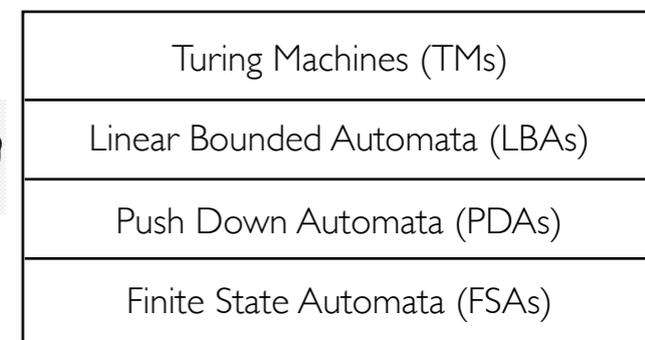


Human Persons (according to Bringsjord)

Human Brains (according to Granger)

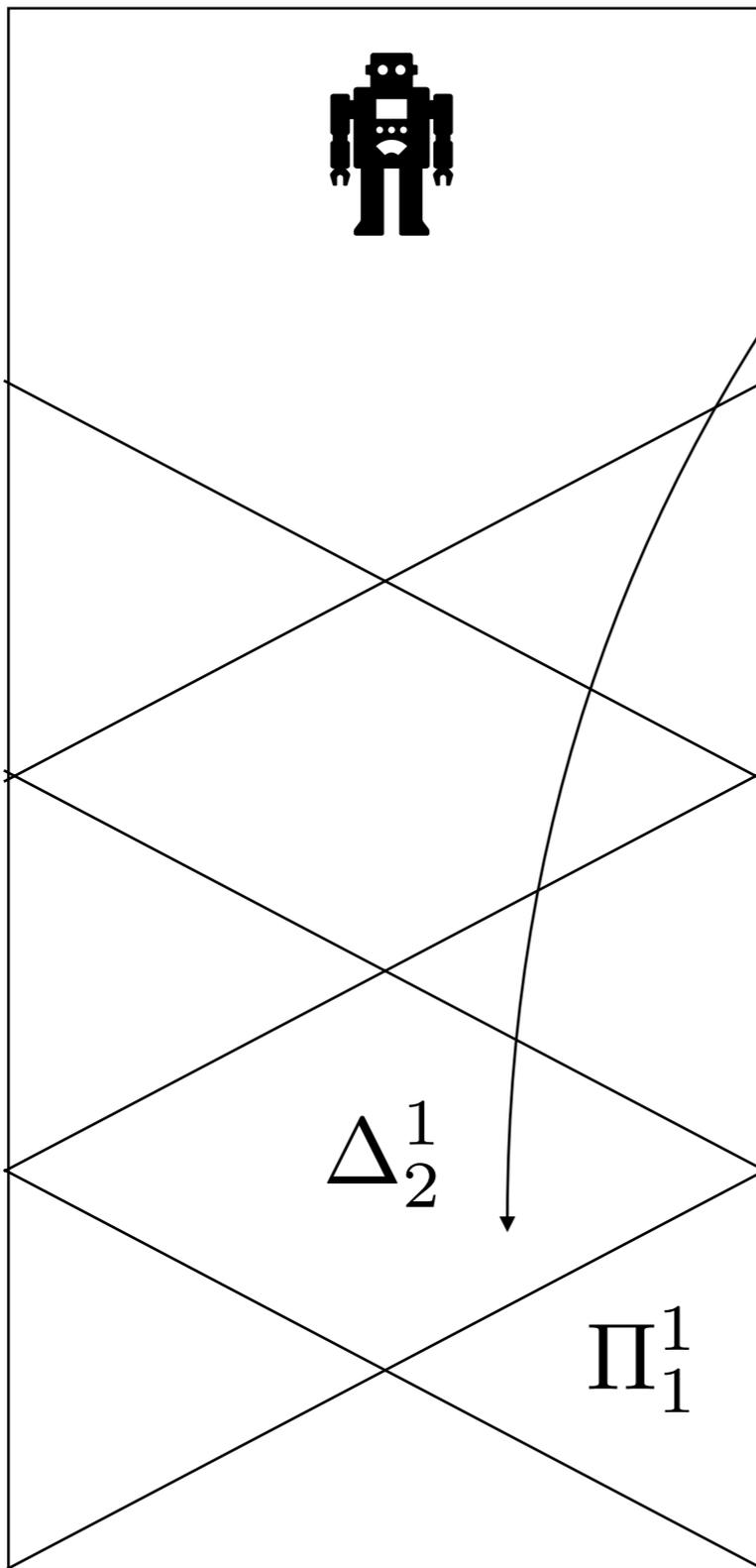


$\mathcal{CH}$  (Chomsky Hierarchy)

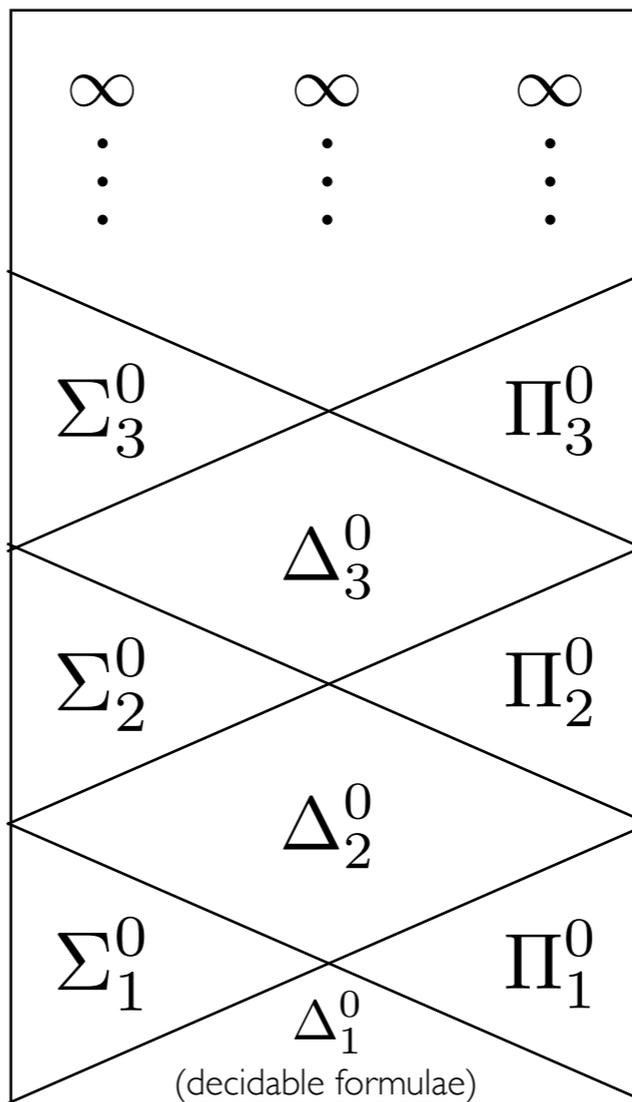


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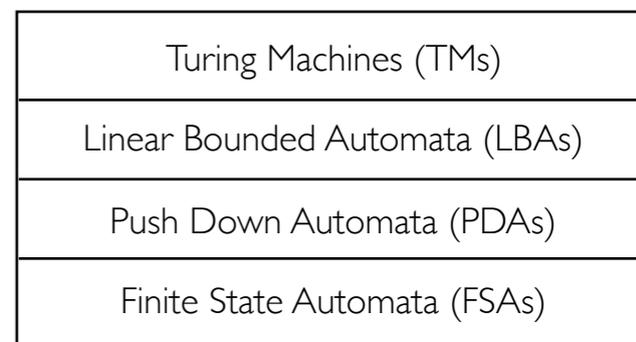
Infinite Time Turing Machines (ITTMs)

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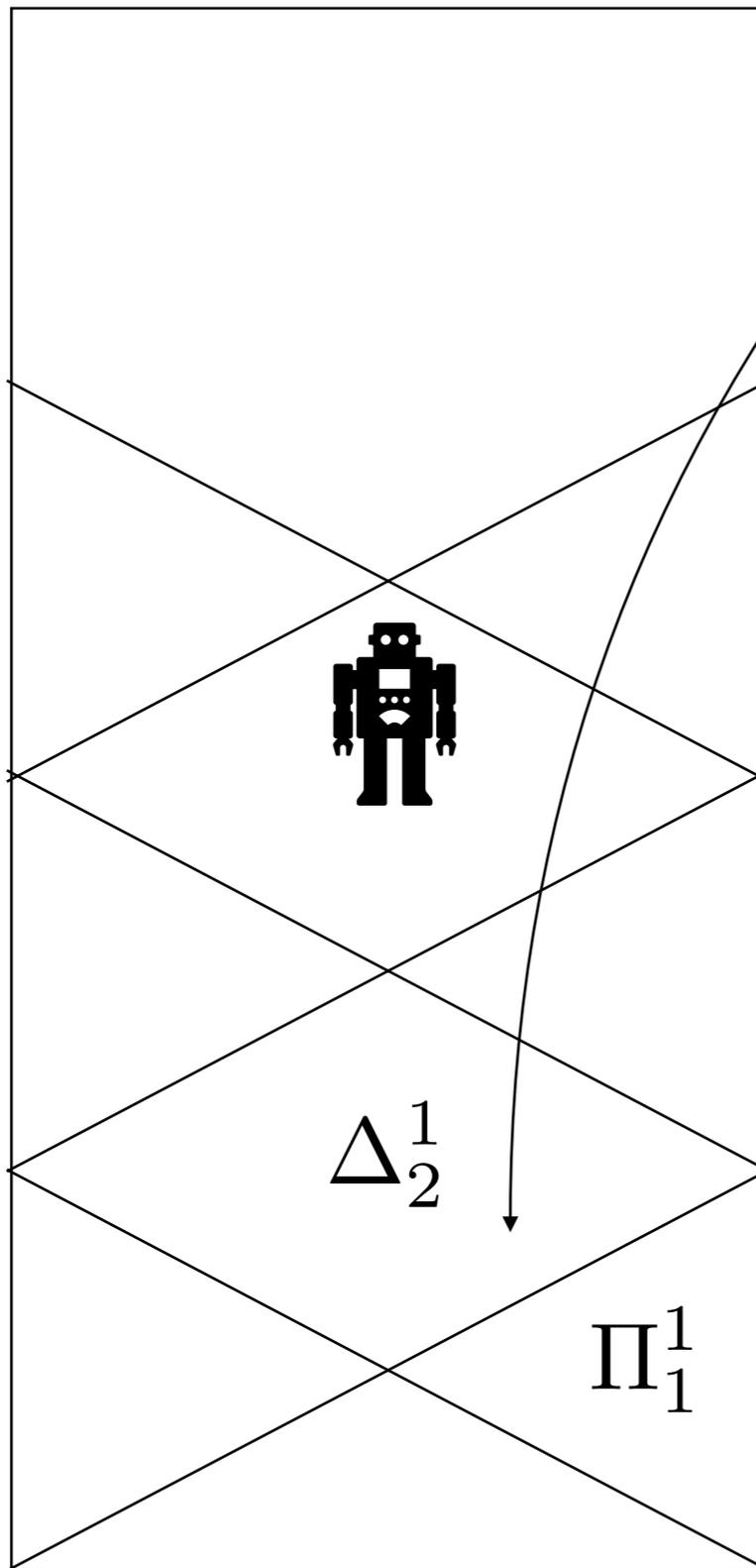
$\mathcal{CH}$  (Chomsky Hierarchy)



$\mathcal{EM}$

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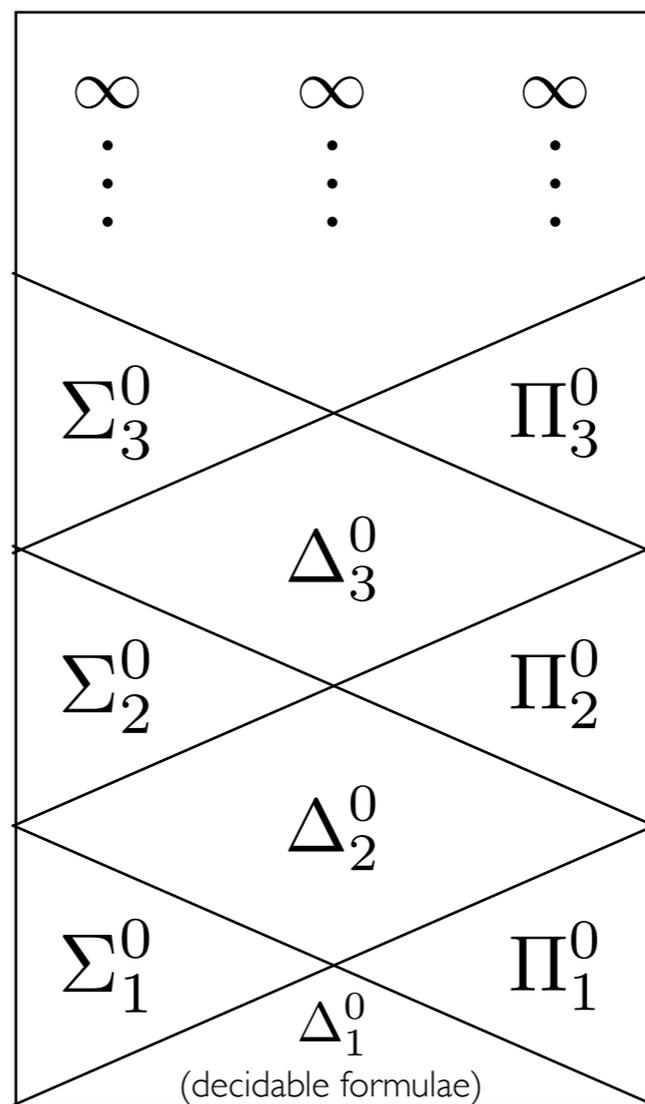
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Infinite Time Turing Machines (ITTMs)

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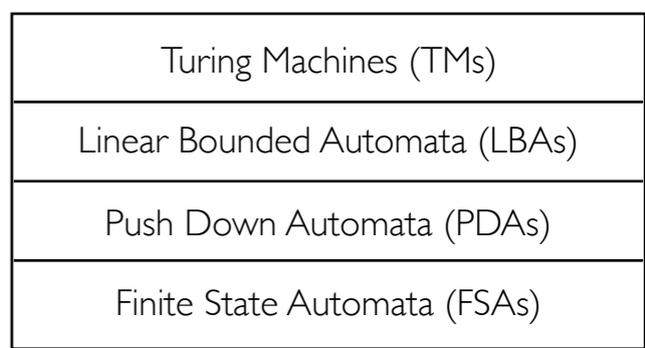
$A^r \mathcal{H}$  (Arithmetic Hierarchy)



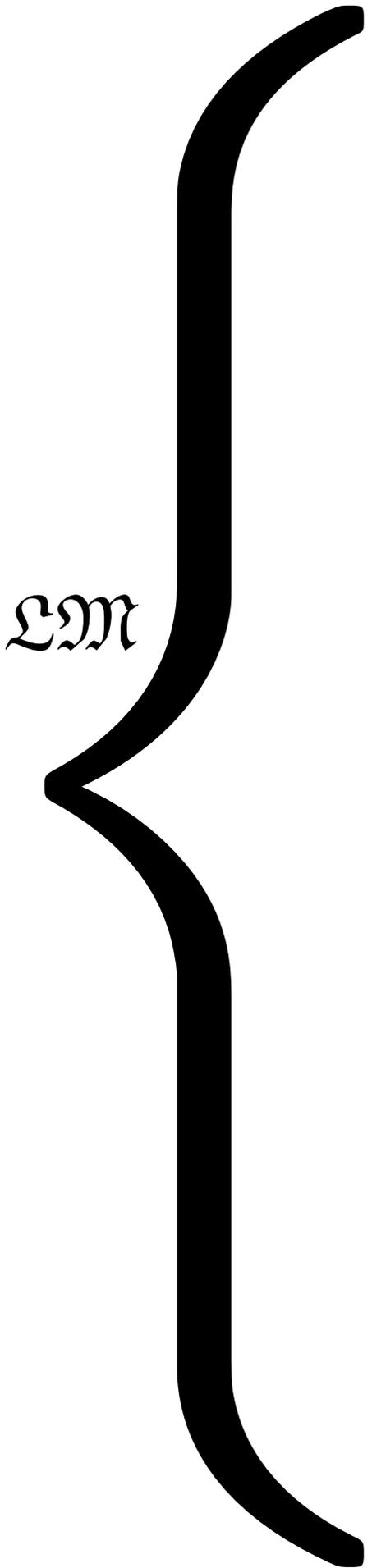
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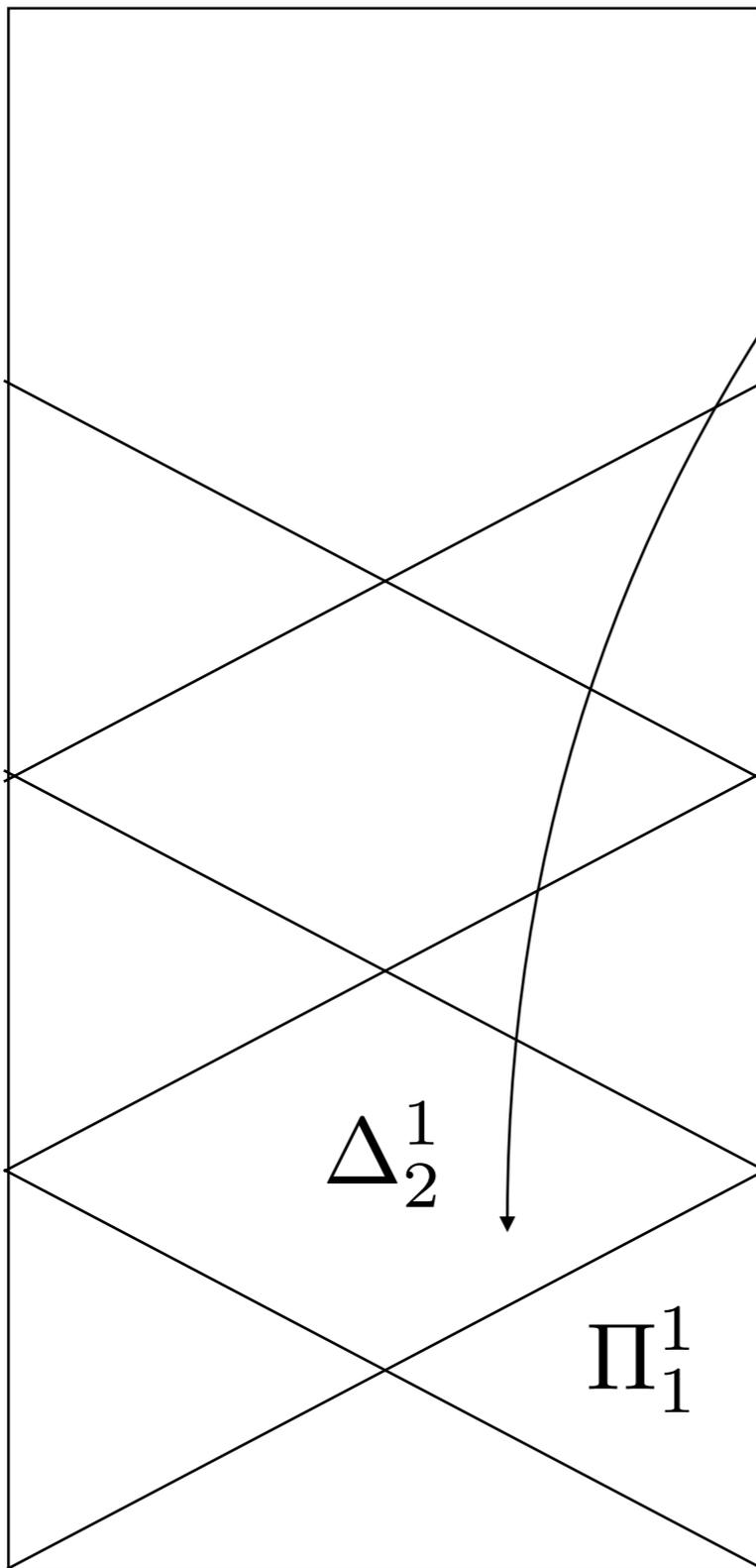


EM



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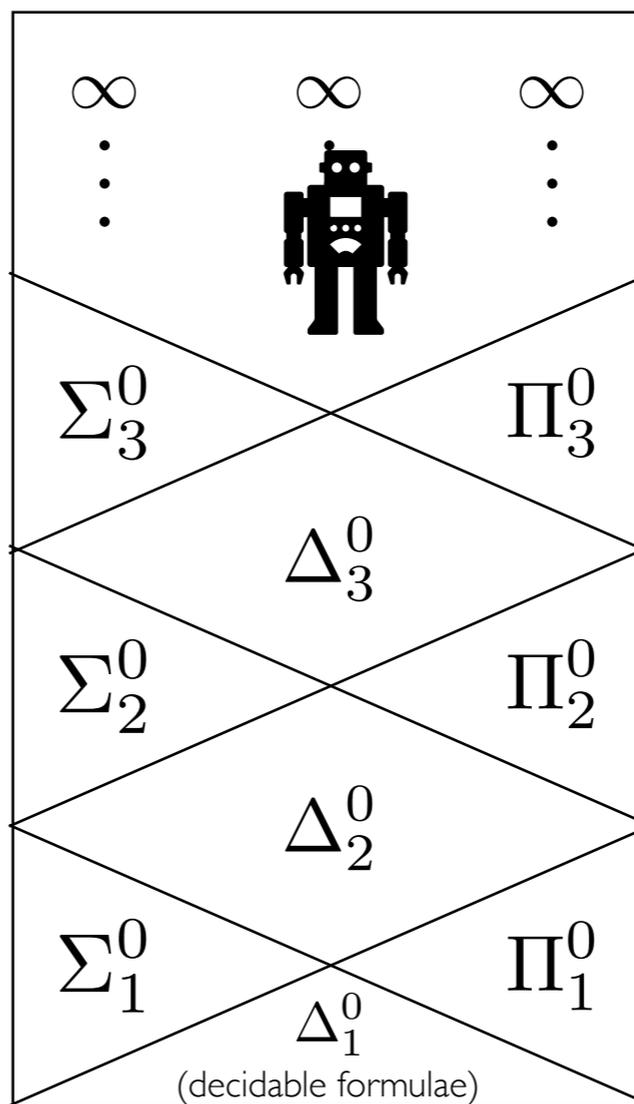
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Infinite Time Turing Machines (ITTMs)

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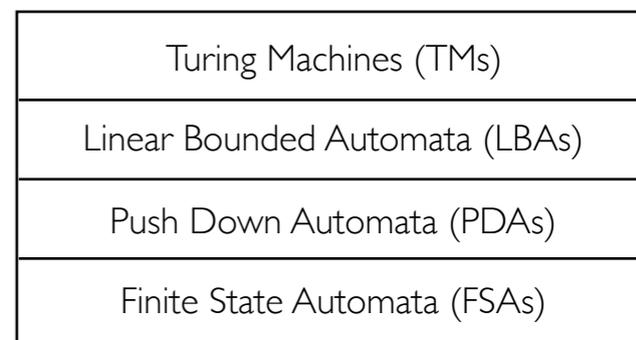
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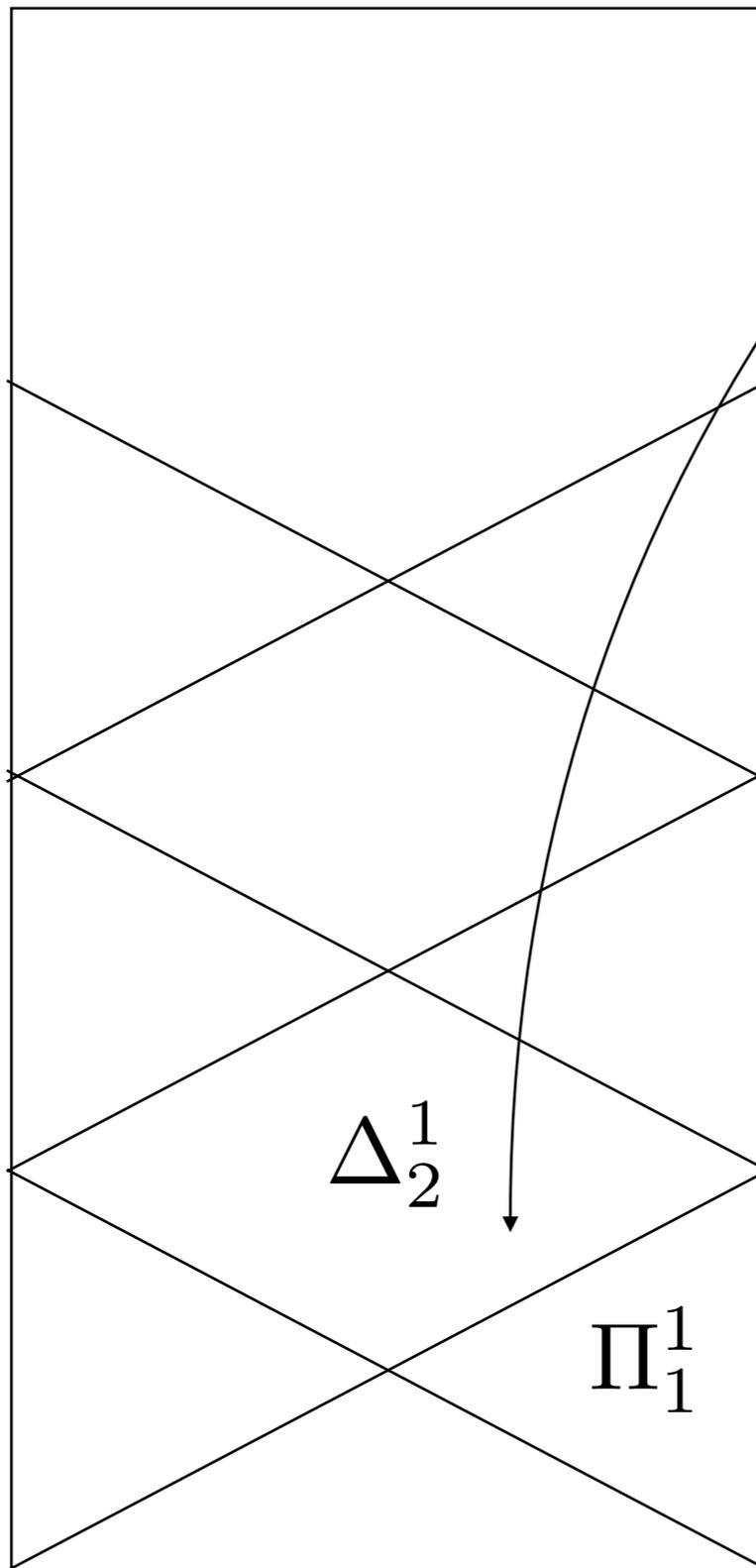
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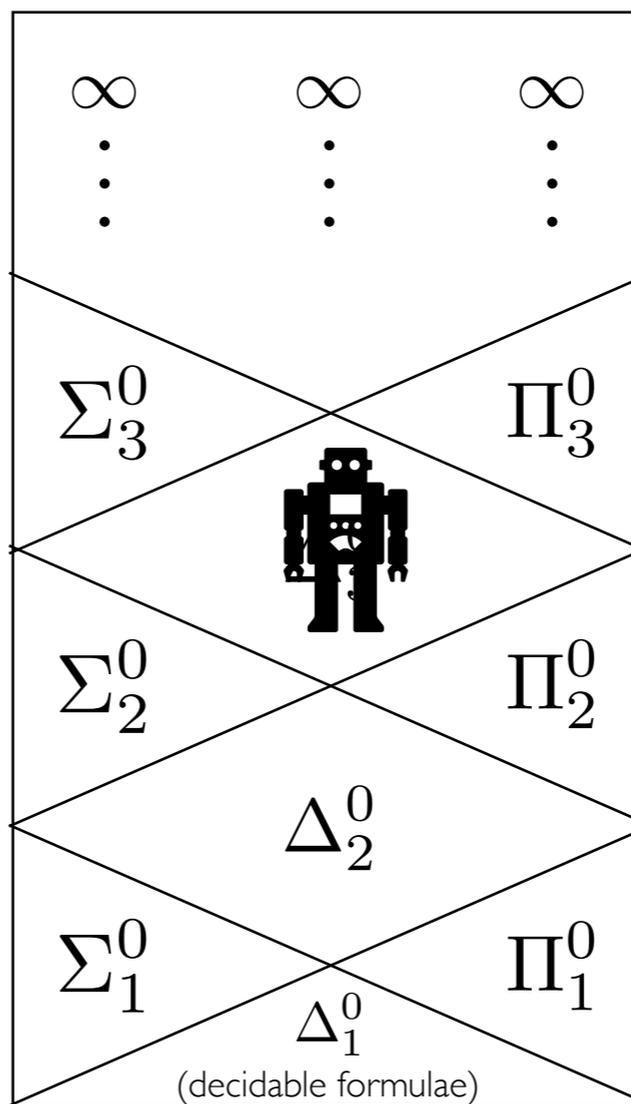
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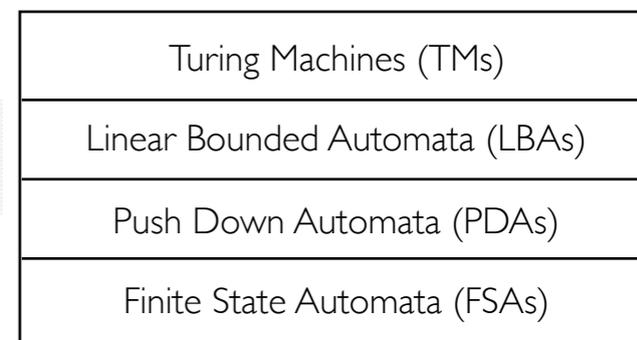


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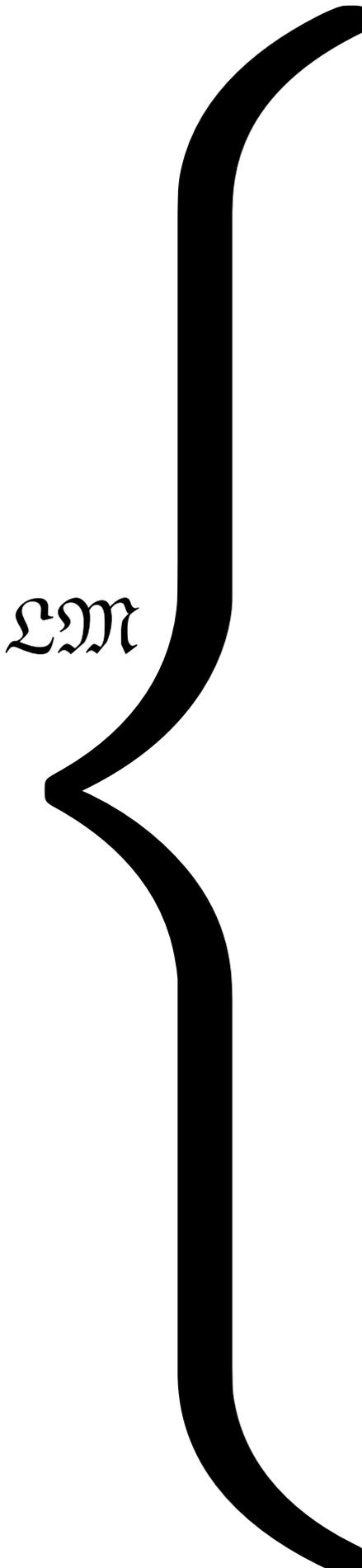
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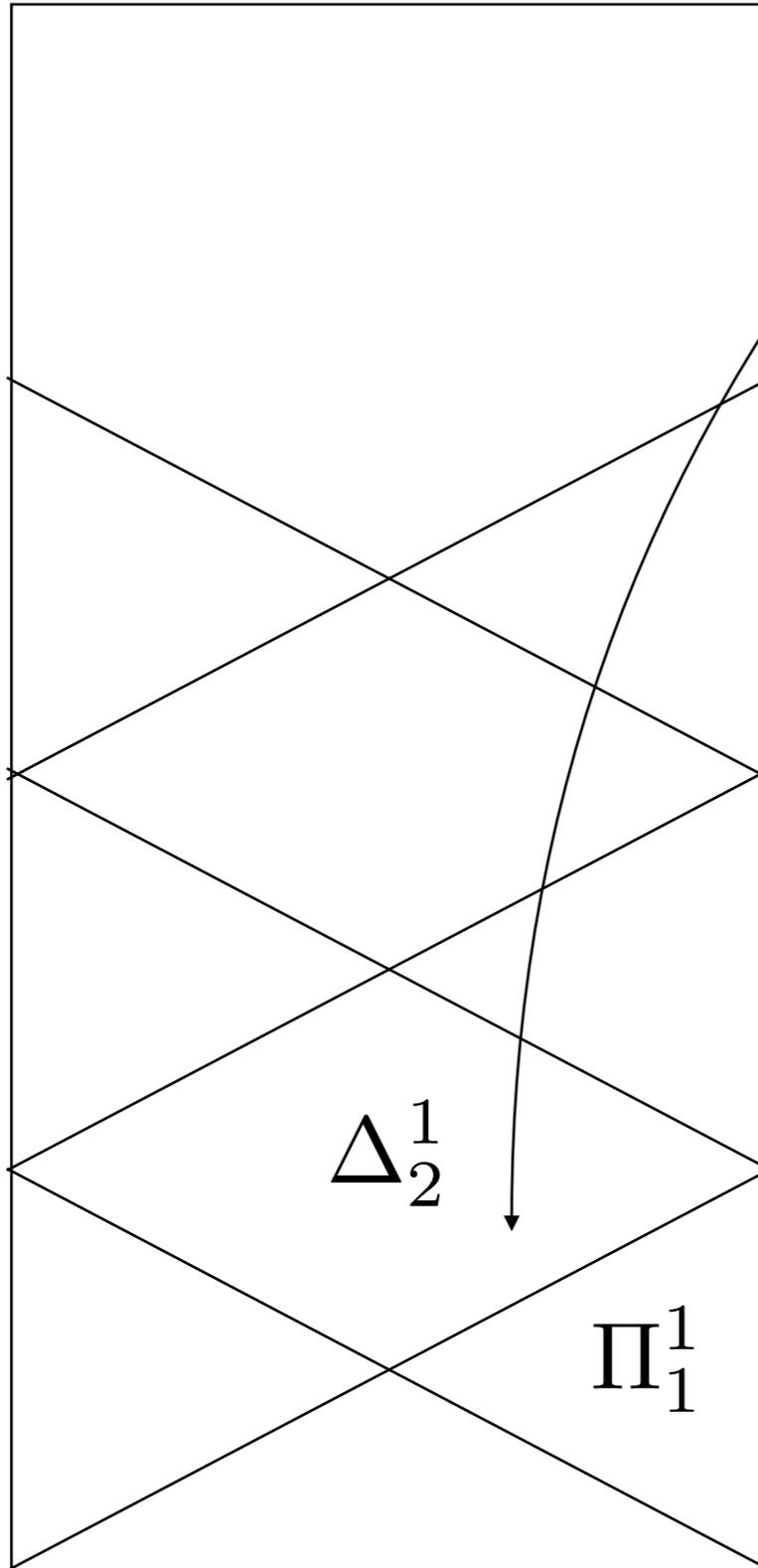


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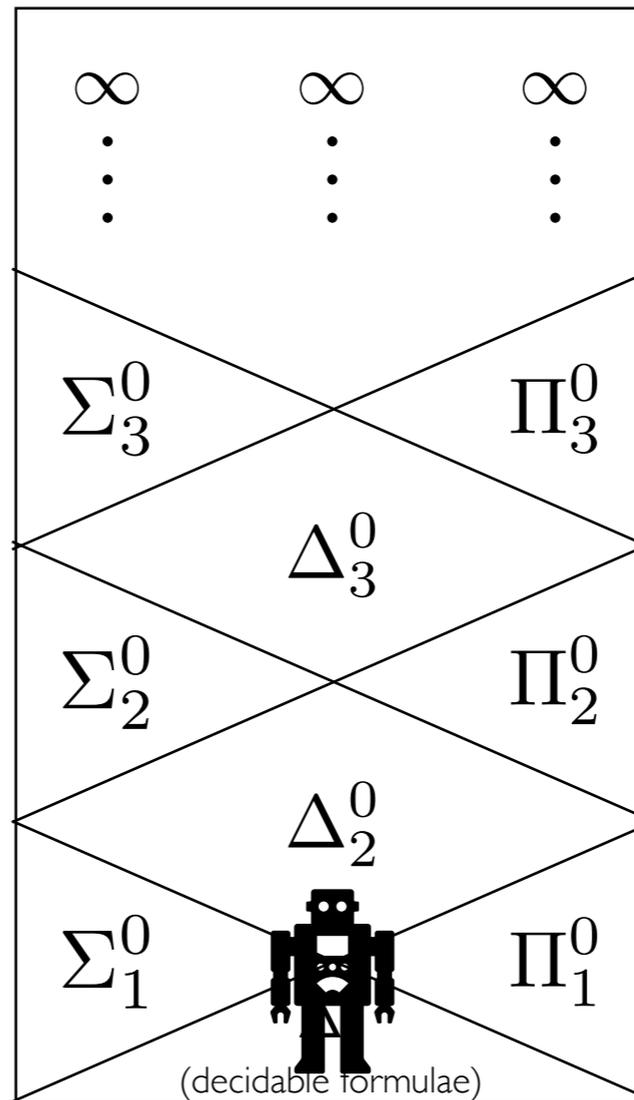
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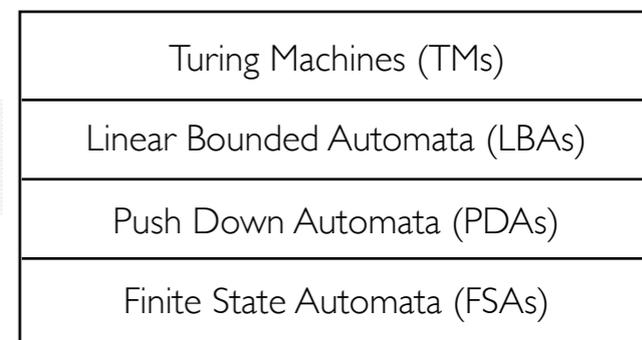
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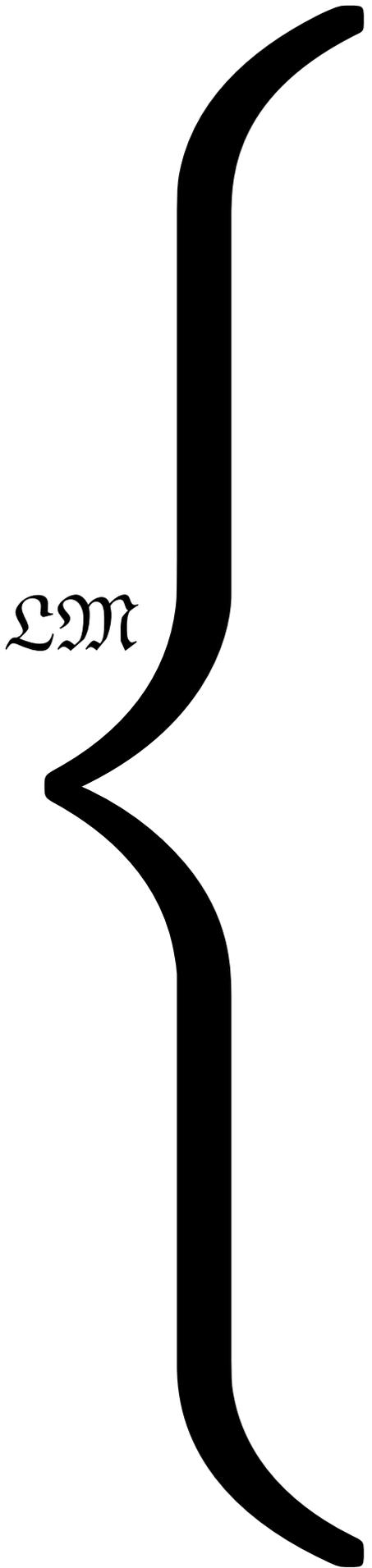
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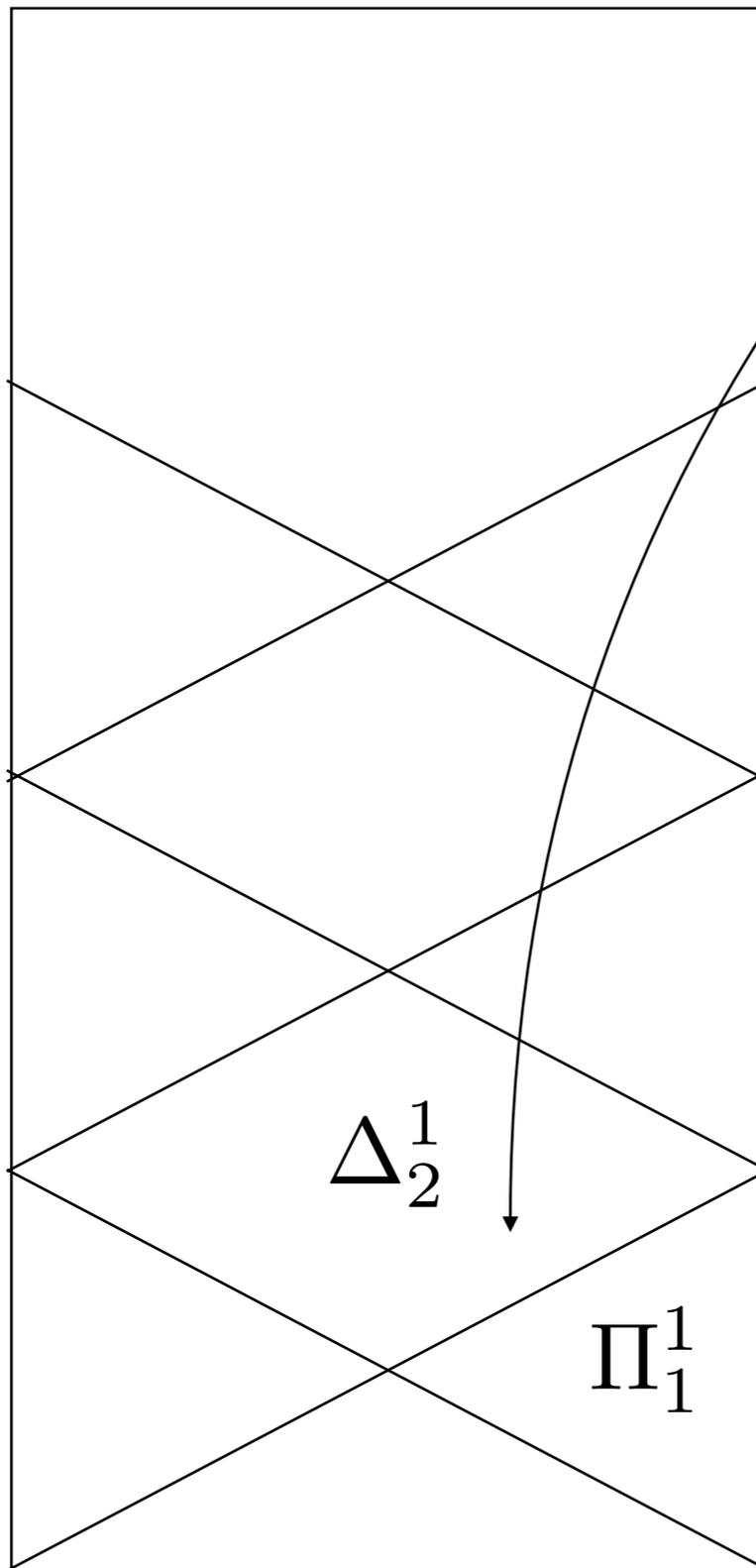


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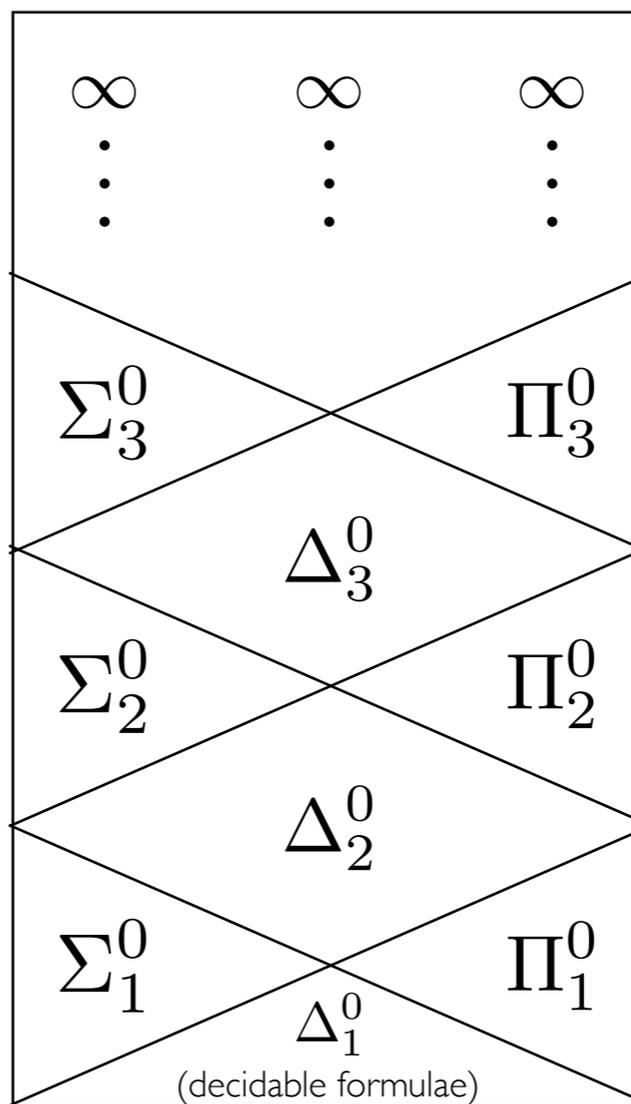
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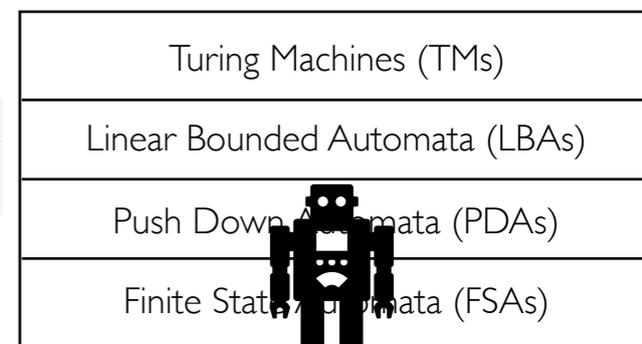
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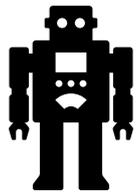


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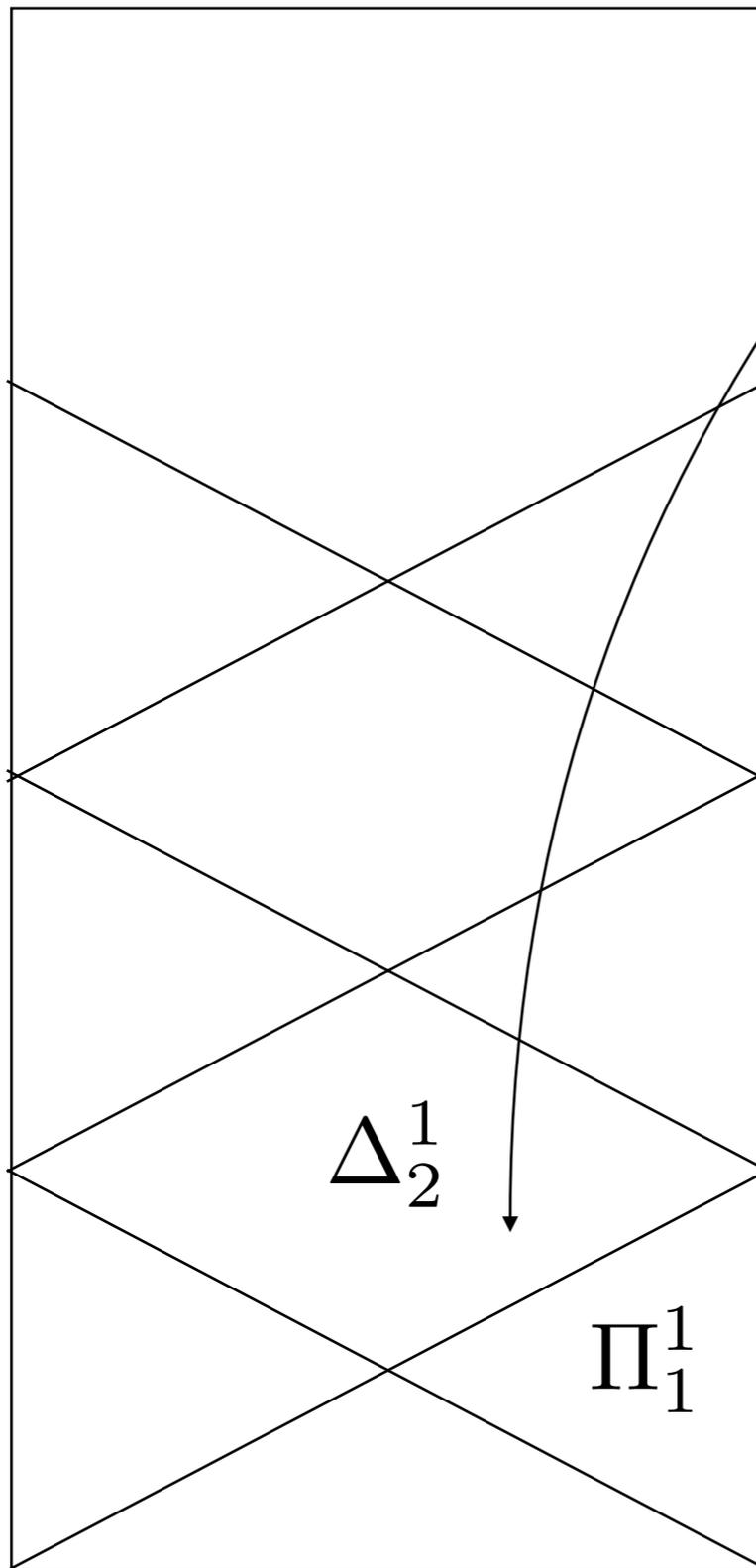


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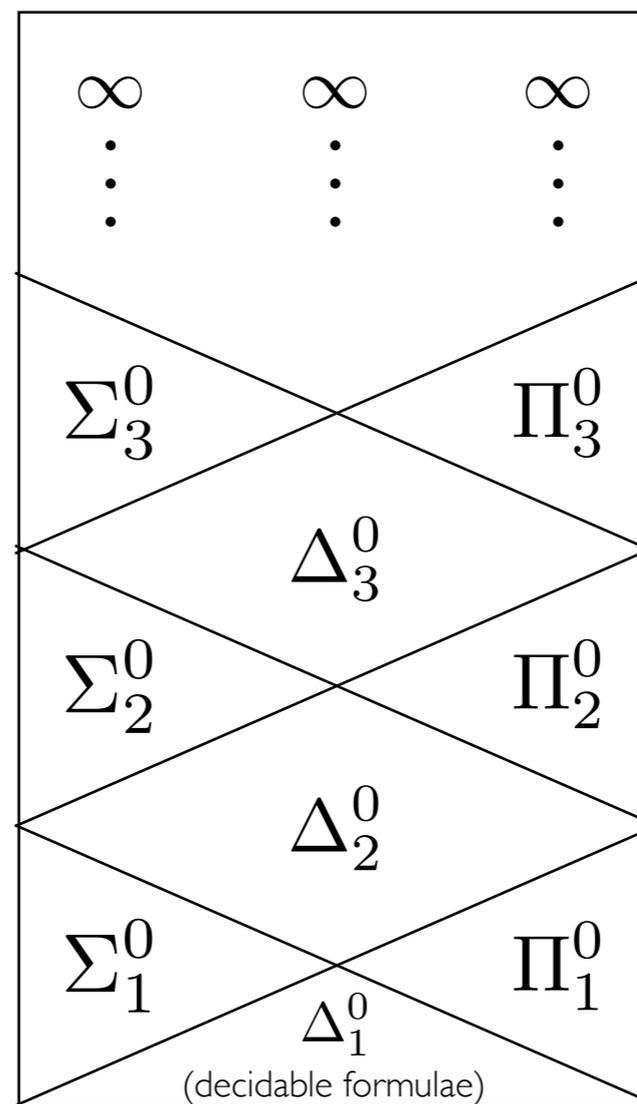


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$A^r \mathcal{H}$  (Arithmetic Hierarchy)

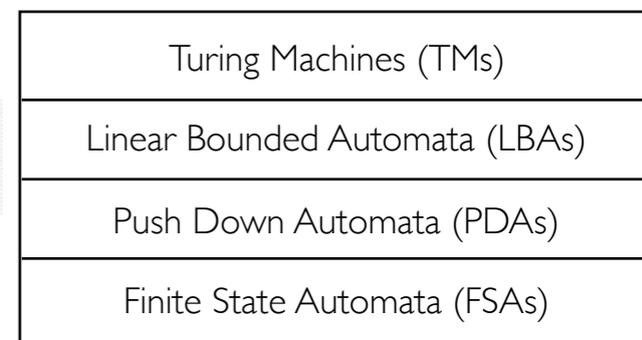


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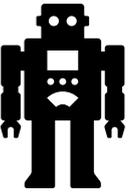
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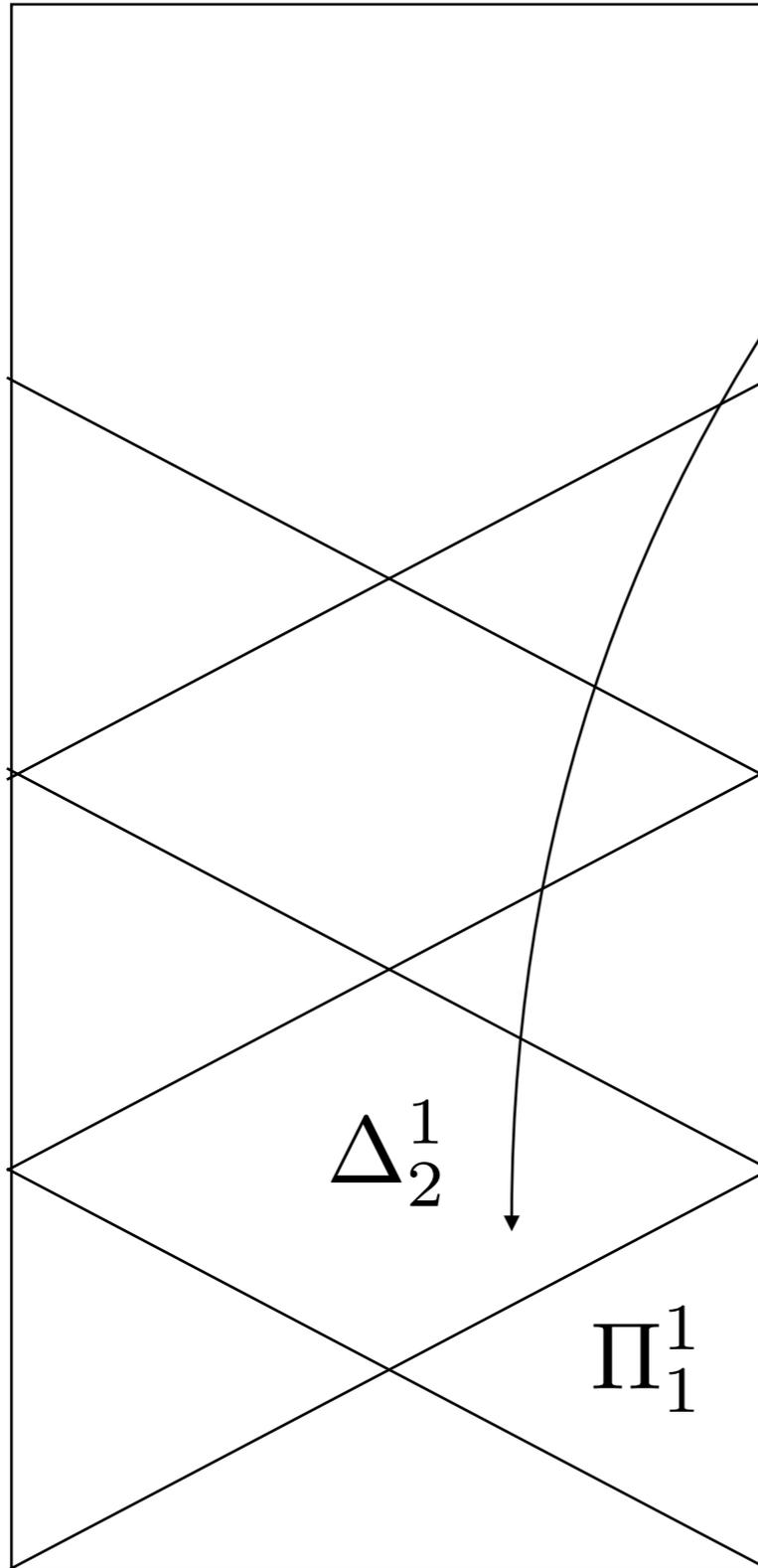
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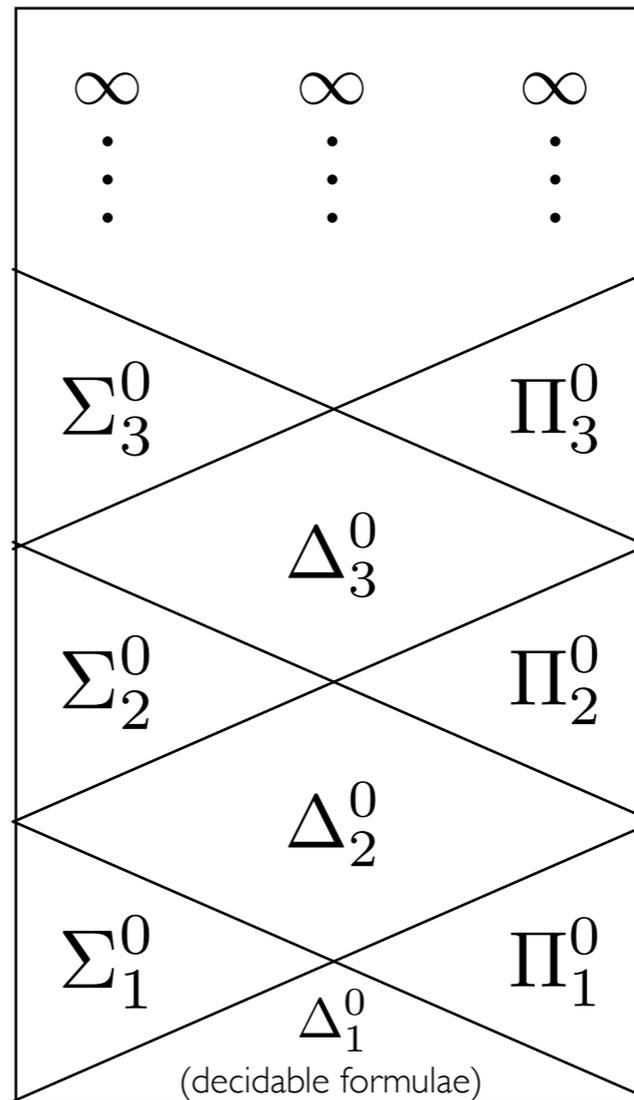


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Infinite Time Turing Machines (ITTTMs)

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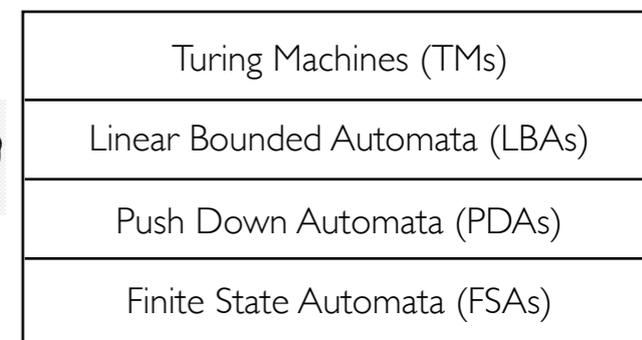


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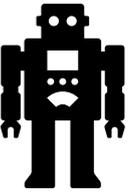


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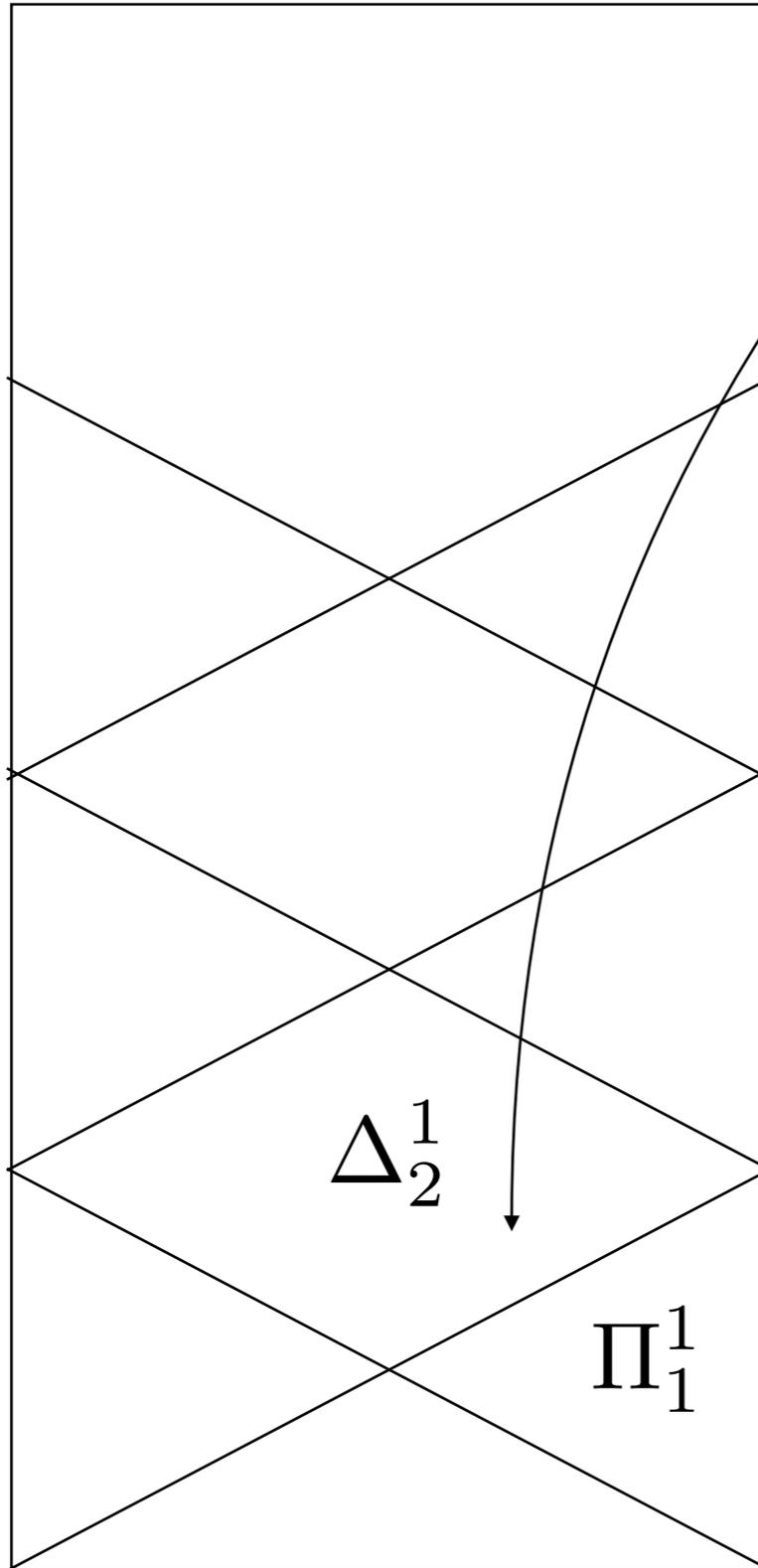


EM

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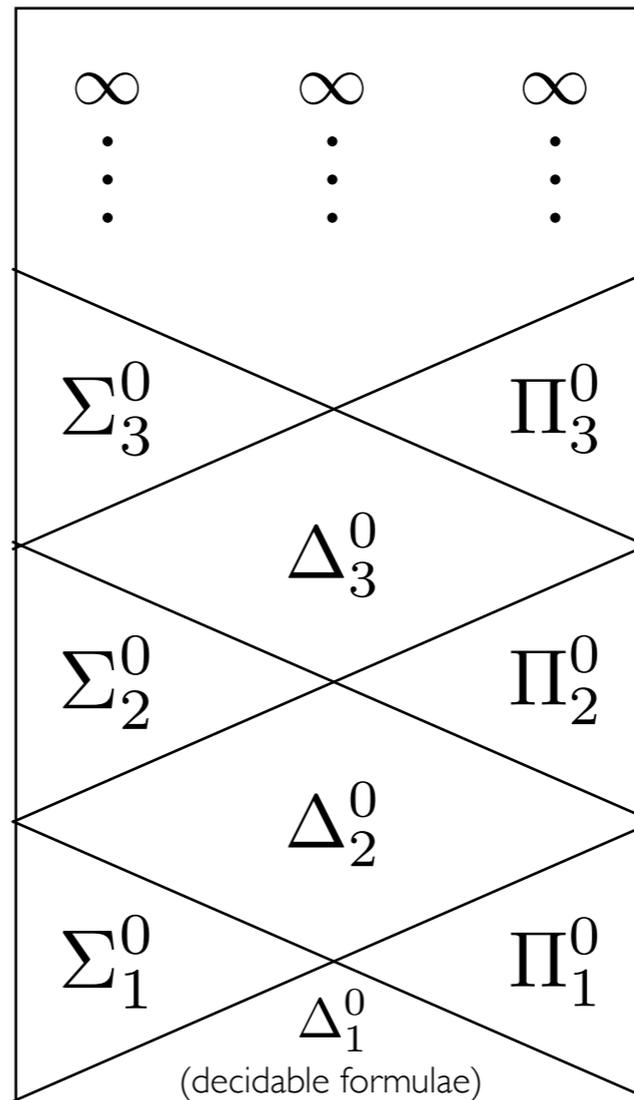


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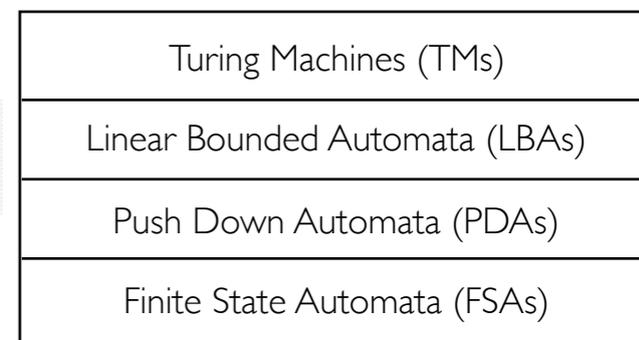


Human Persons (according to Bringsjord)

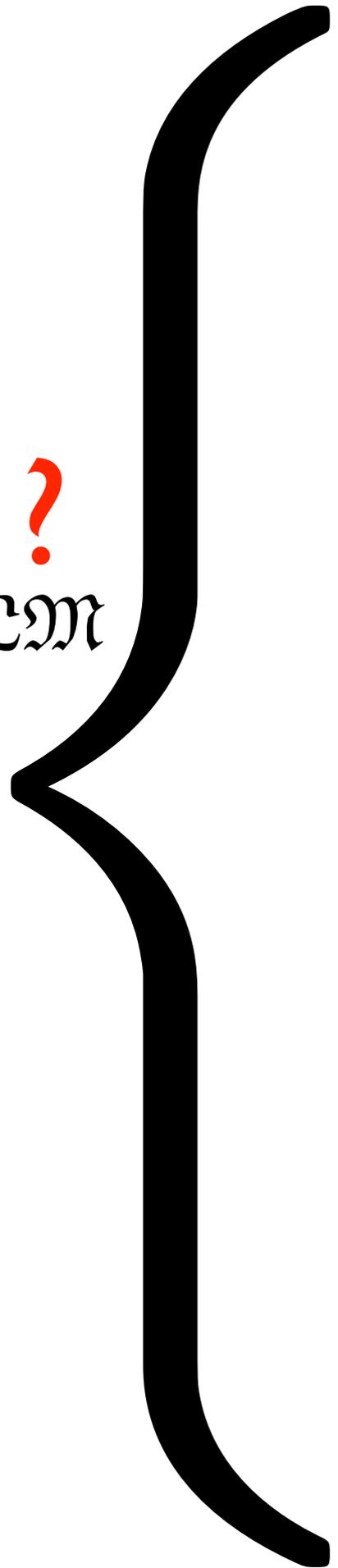
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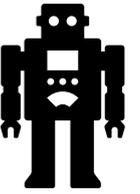
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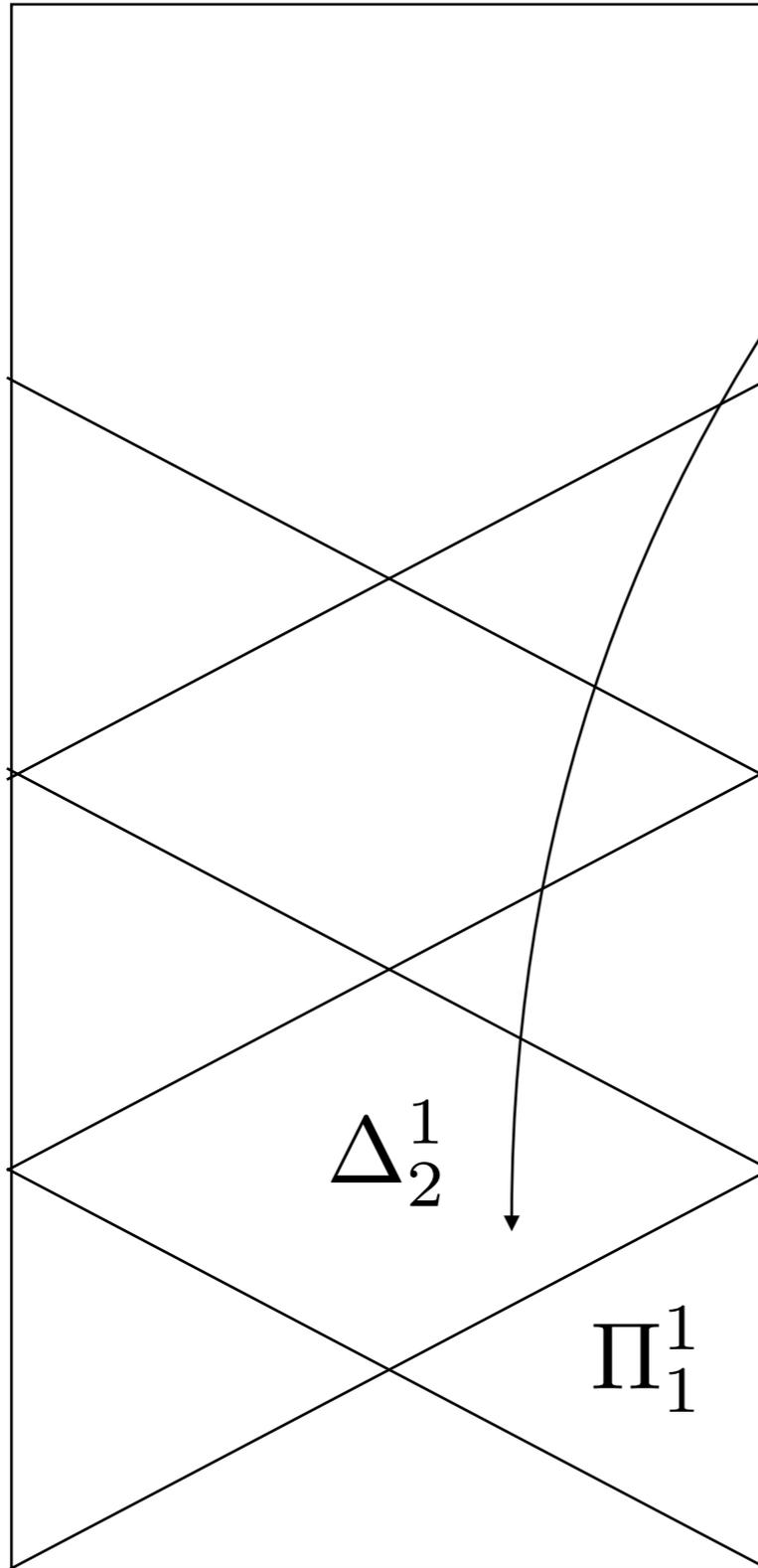
?  
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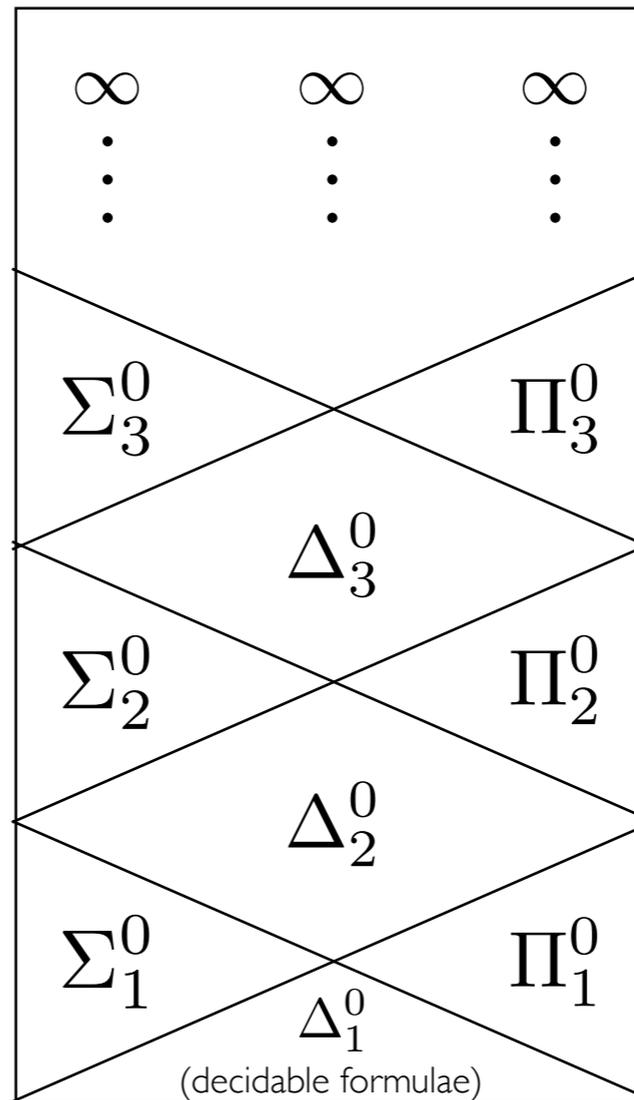


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Infinite Time Turing Machines (ITTMs)

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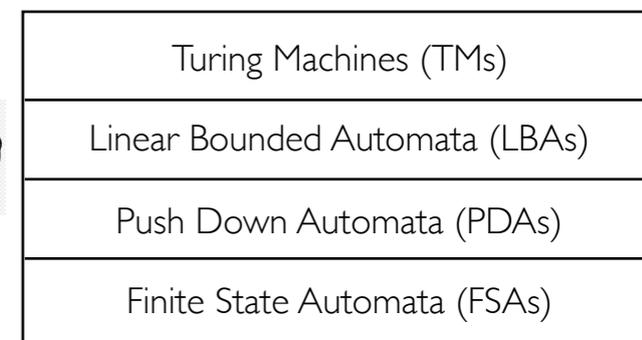


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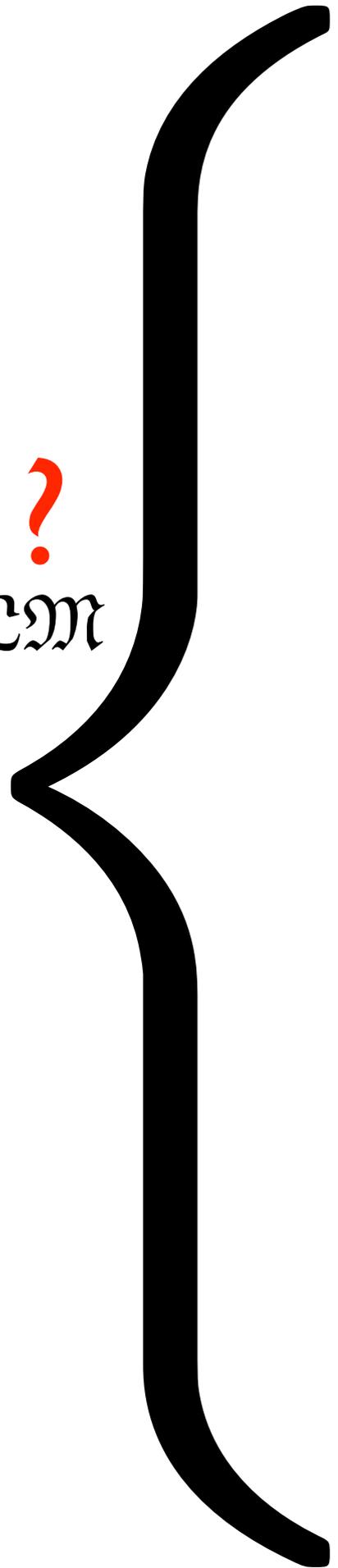
Human Brains (according to Granger)



$\mathcal{CH}$  (Chomsky Hierarchy)

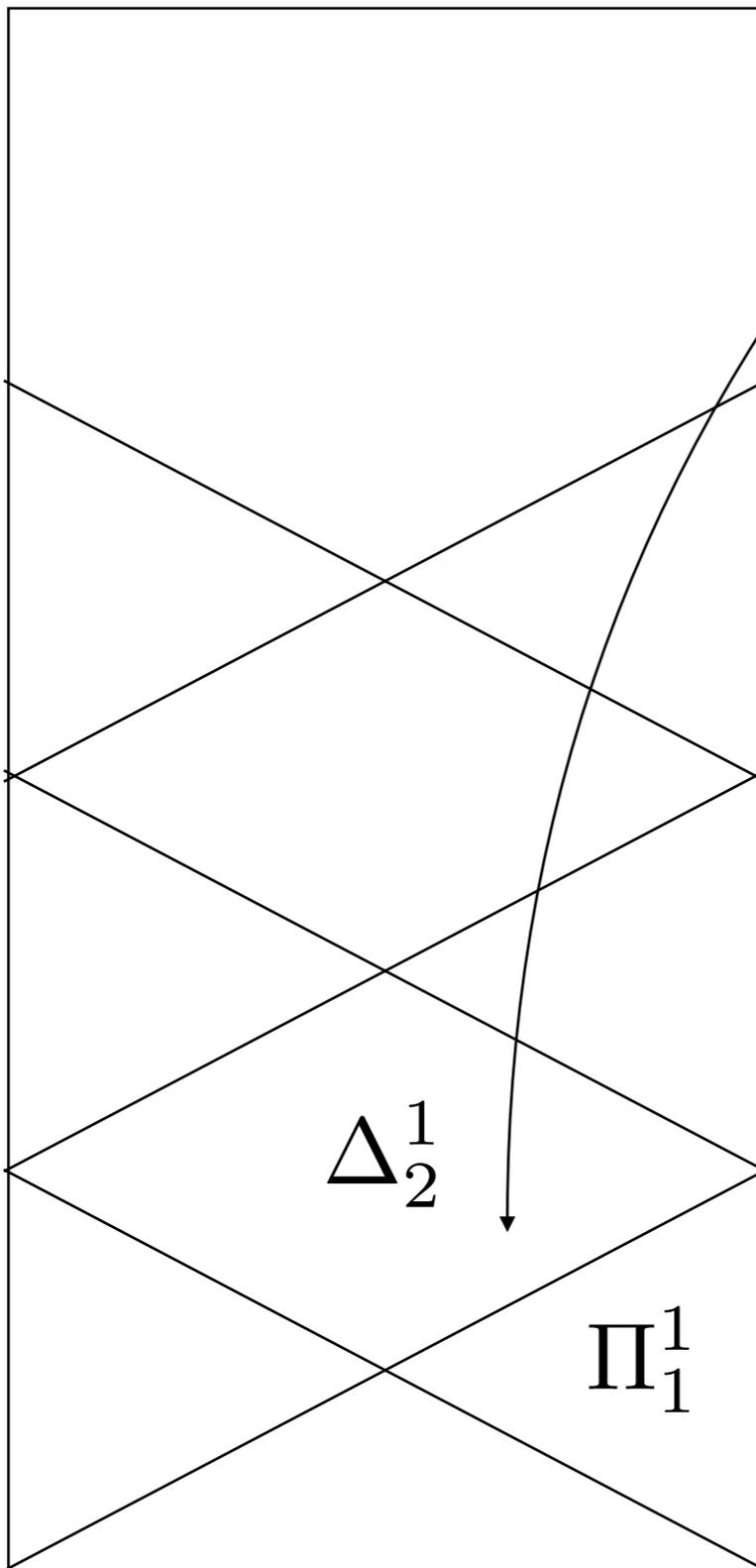


?  
 $\mathcal{EM}$



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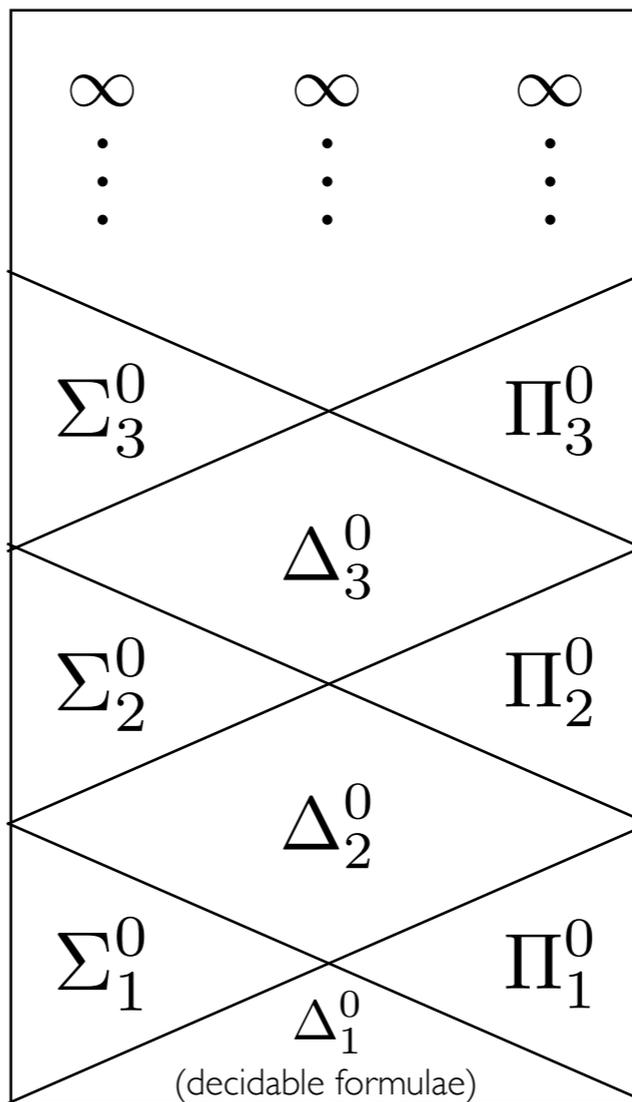
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Infinite Time Turing Machines (ITTMs)

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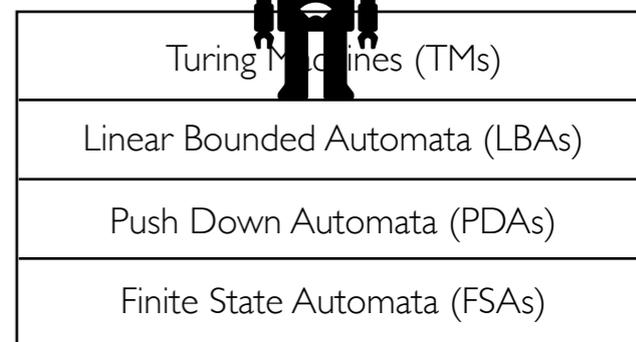
$A^r \mathcal{H}$  (Arithmetic Hierarchy)



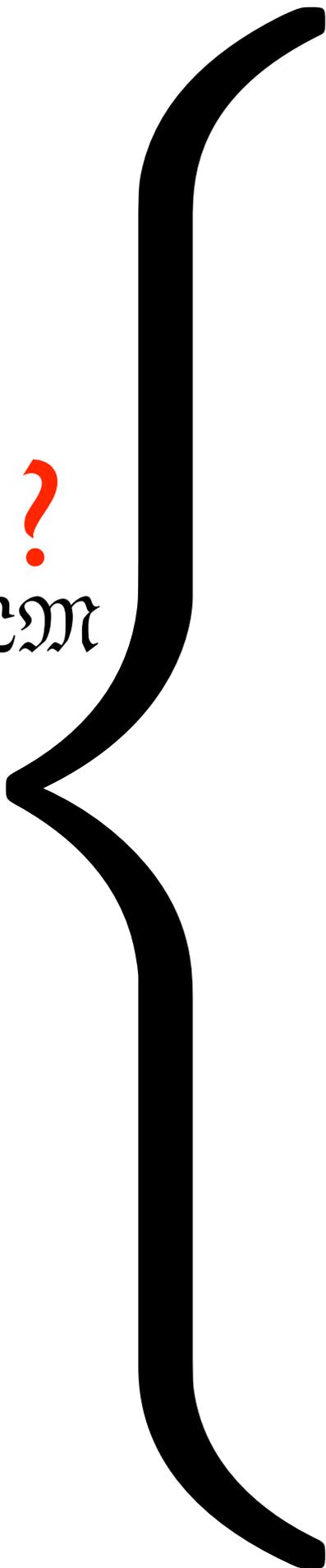
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$\mathcal{CH}$  (Churchs Hierarchy)

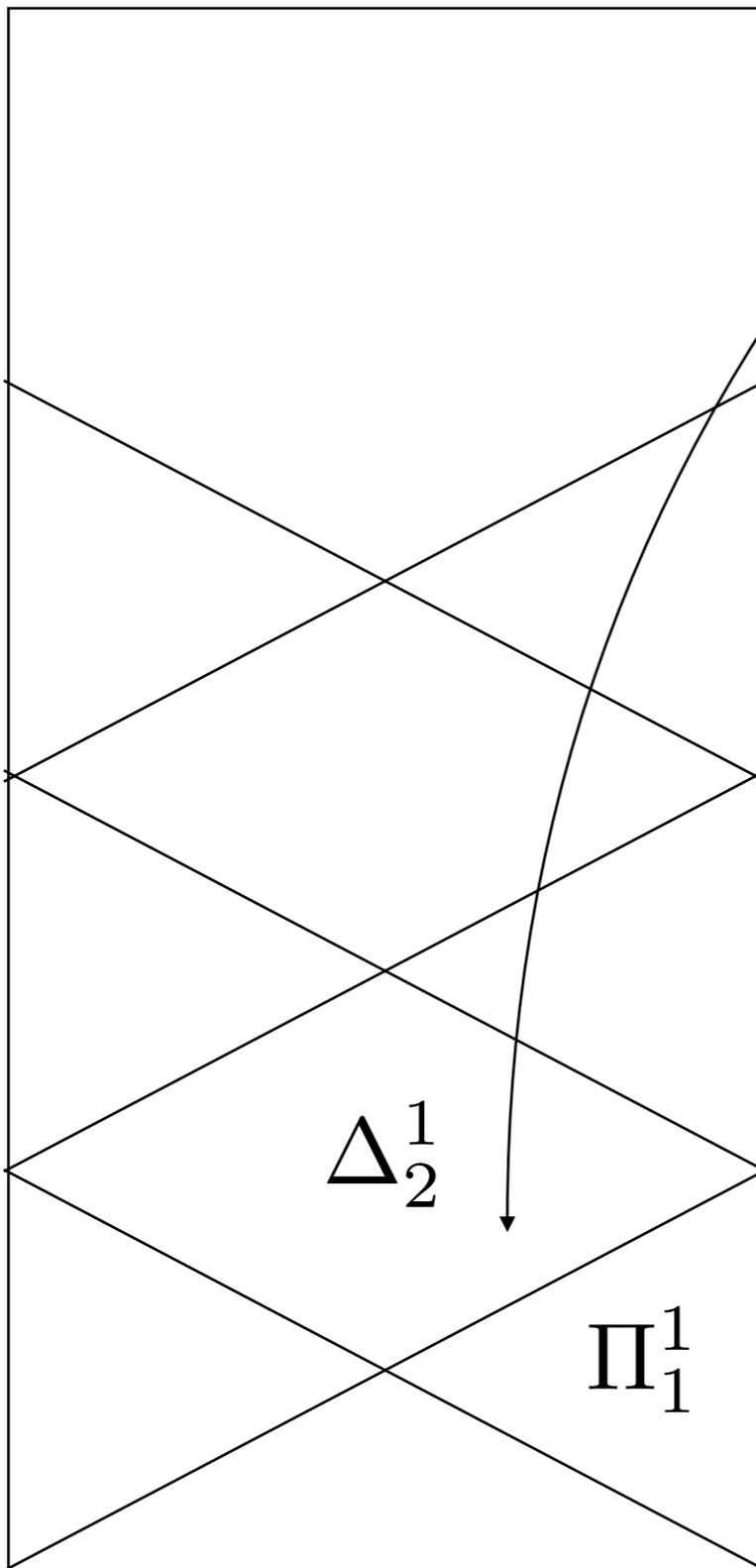


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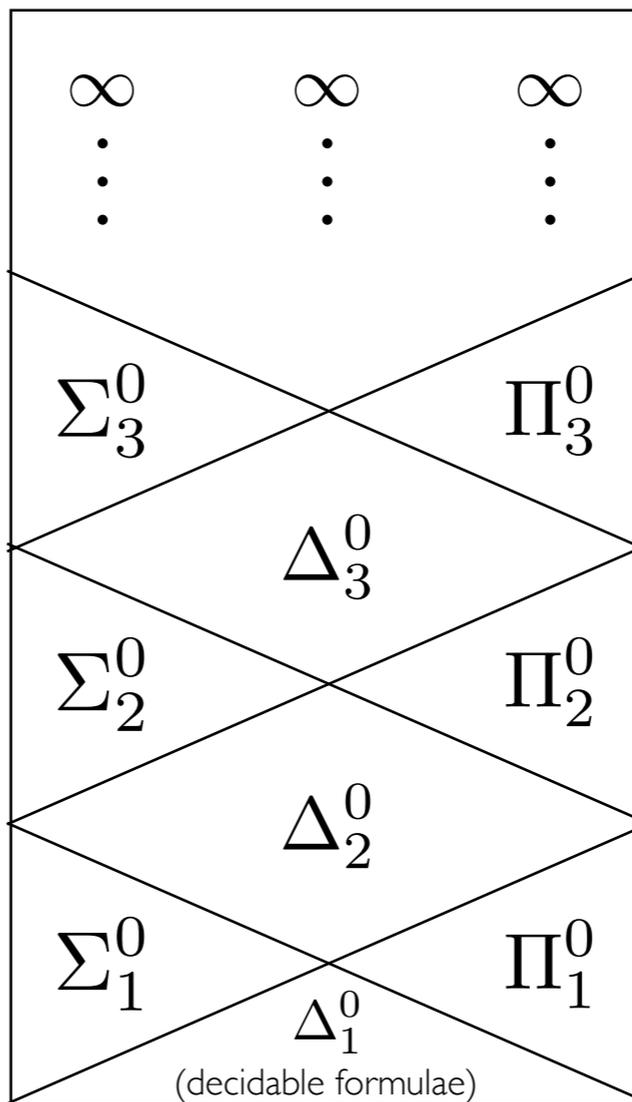
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Infinite Time Turing Machines (ITTTMs)

Human Persons  
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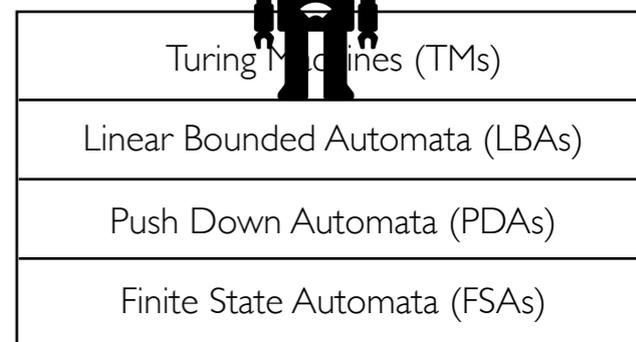
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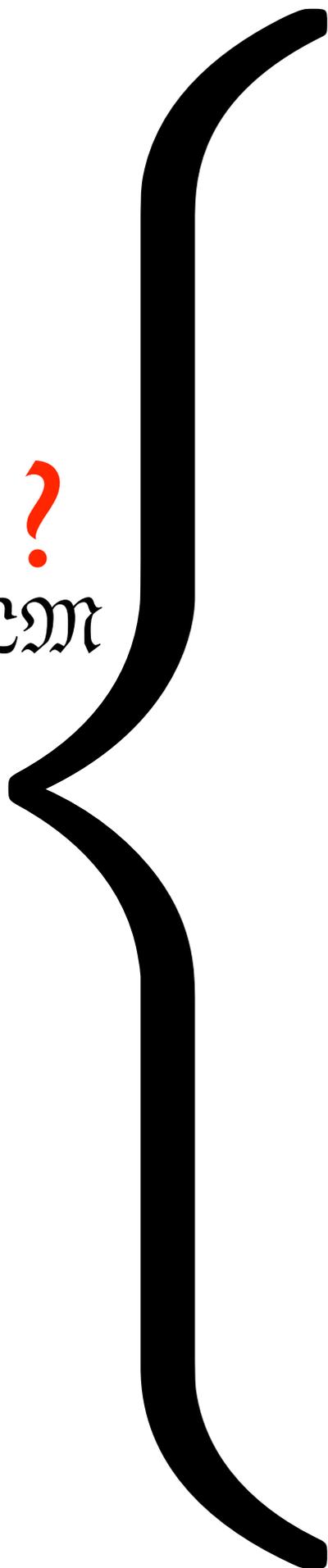
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$\mathcal{CH}$  (Computational Hierarchy)

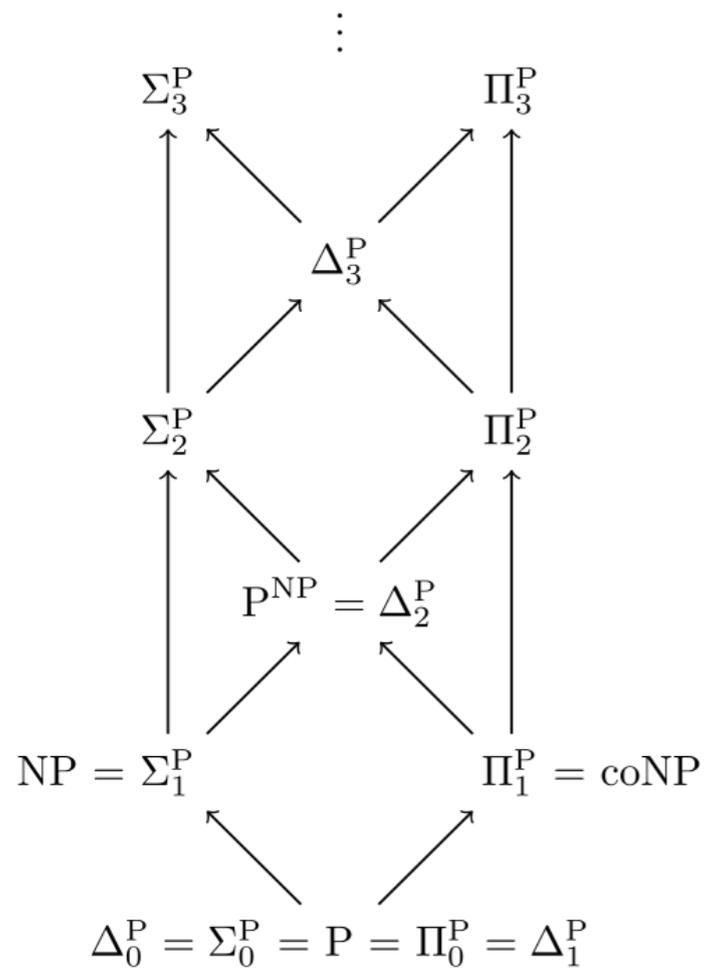
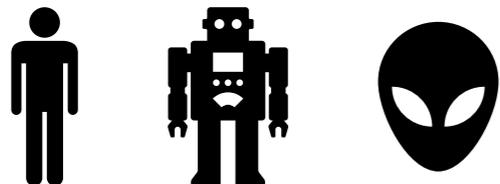


EM ?



# Polynomial Hierarchy, Part I

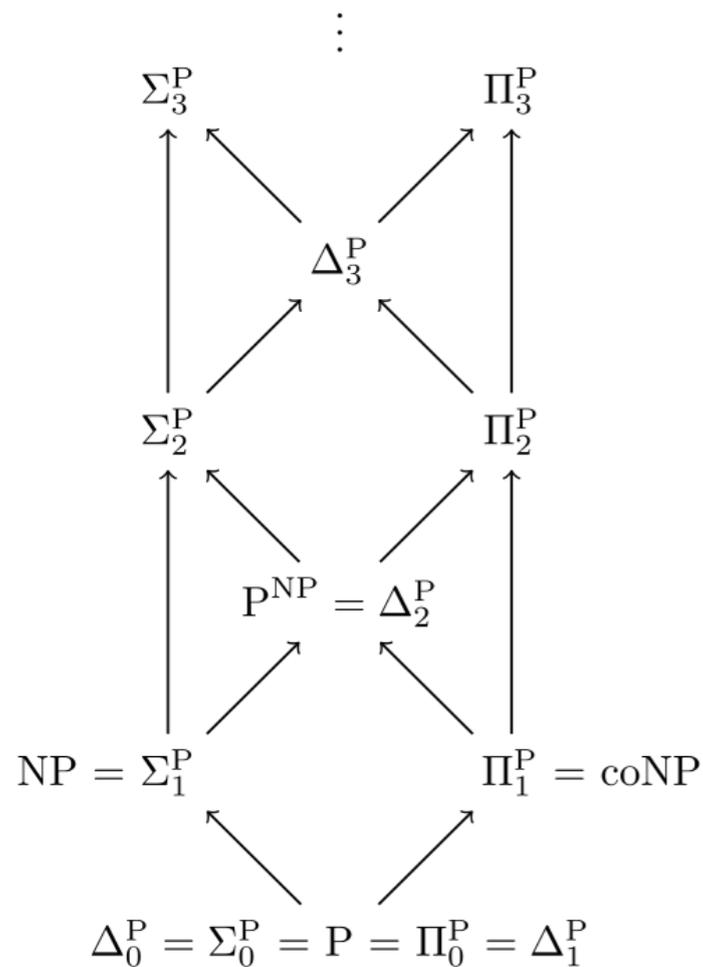
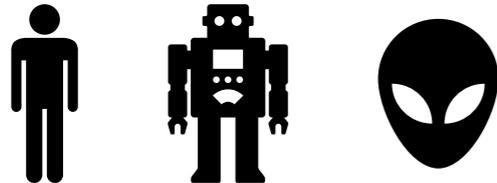
(via formal logic, directly; a start)



# Polynomial Hierarchy, Part I

(via formal logic, directly; a start)

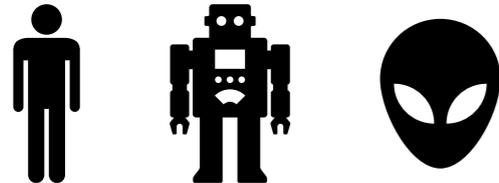
We say that a relation  $R(u, y_1, \dots, y_n)$  is polytime iff there is a deterministic Turing Machine  $m$  and a polynomial  $p$  s.t.  $m$  decides this relation in  $p(|u|)$ .



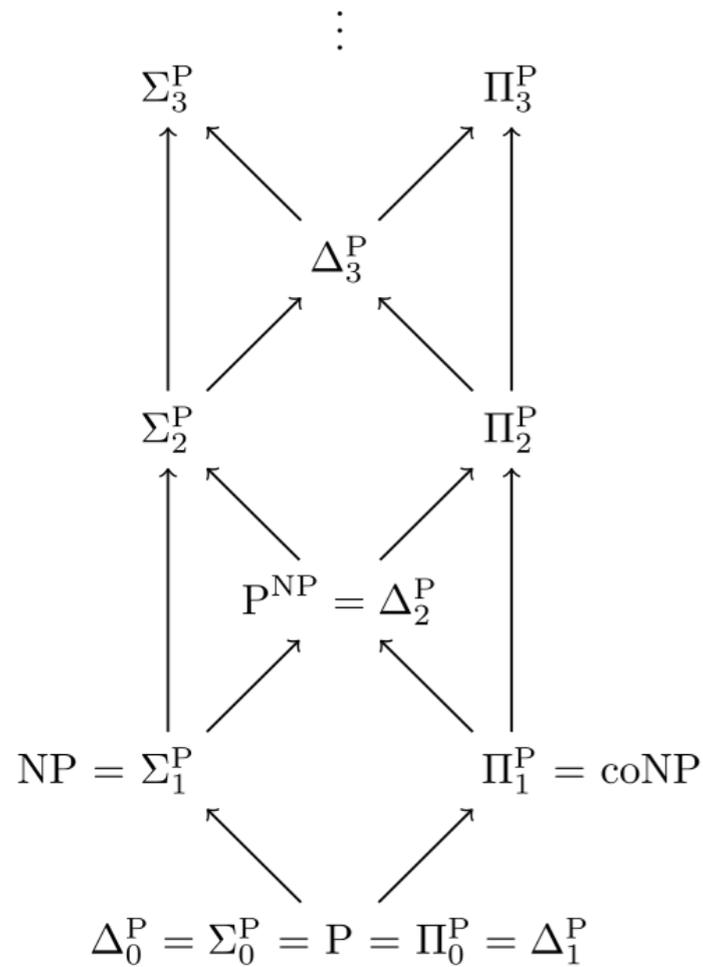
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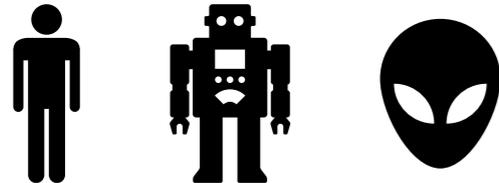


$L \in \mathbf{NP}$  iff: there's a polytime relation  $R$  s.t.  $u \in L$  iff  $\exists y R(u, y)$ .



# Polynomial Hierarchy, Part I

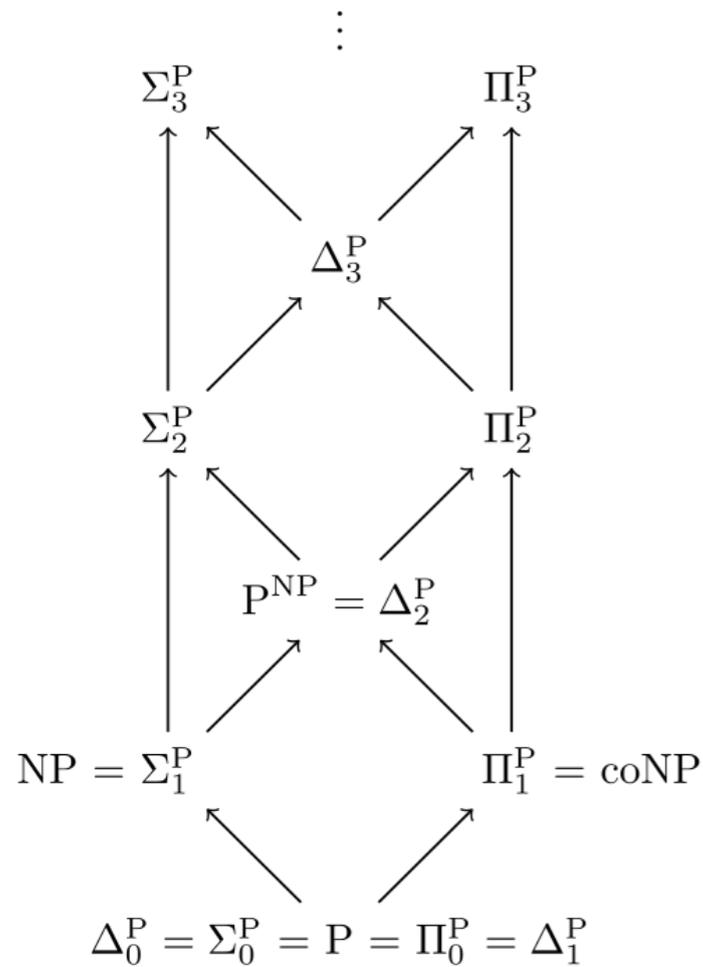
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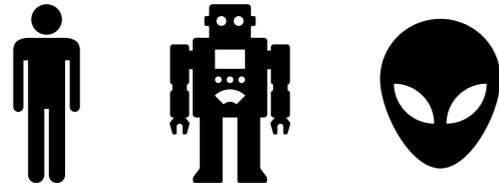
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E.g.: We can prove  $\mathbf{SAT} \in \mathbf{NP}$  because we have a polytime relation  $R$  s.t.  $\phi \in \mathbf{SAT}$  iff  $\exists y R(\phi \in \mathcal{L}_{pc}, \langle \text{assignments to Boolean vars} \rangle)$ , where these assignments produce truth.



# Polynomial Hierarchy, Part I

(via formal logic, directly; a start)

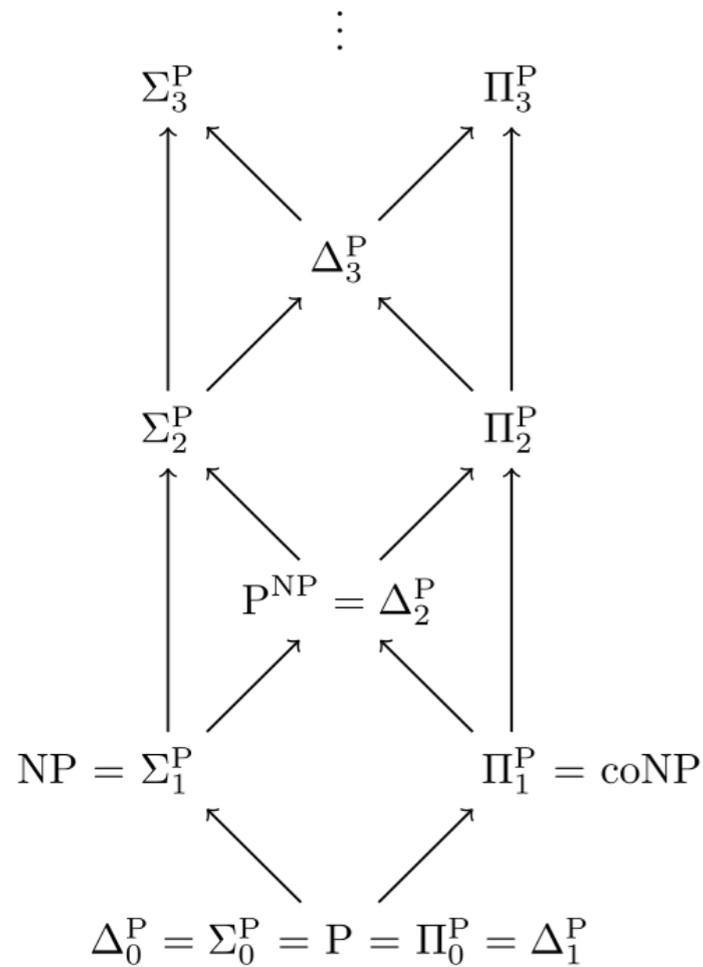


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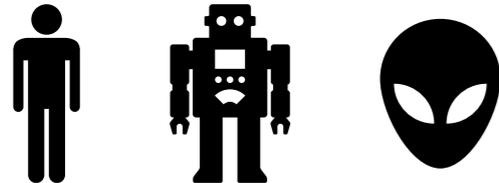
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$L \in \mathbf{coNP}$  iff: there's a polytime relation  $R$  s.t.  $u \in L$  iff  $\forall yR(u, y)$ .



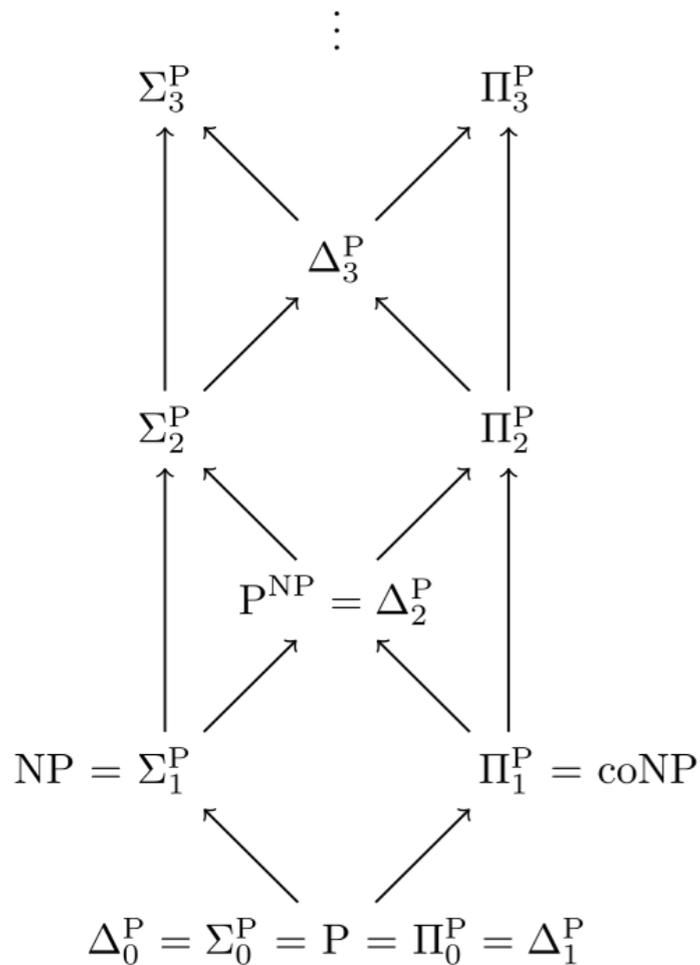
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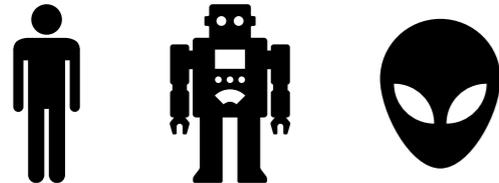
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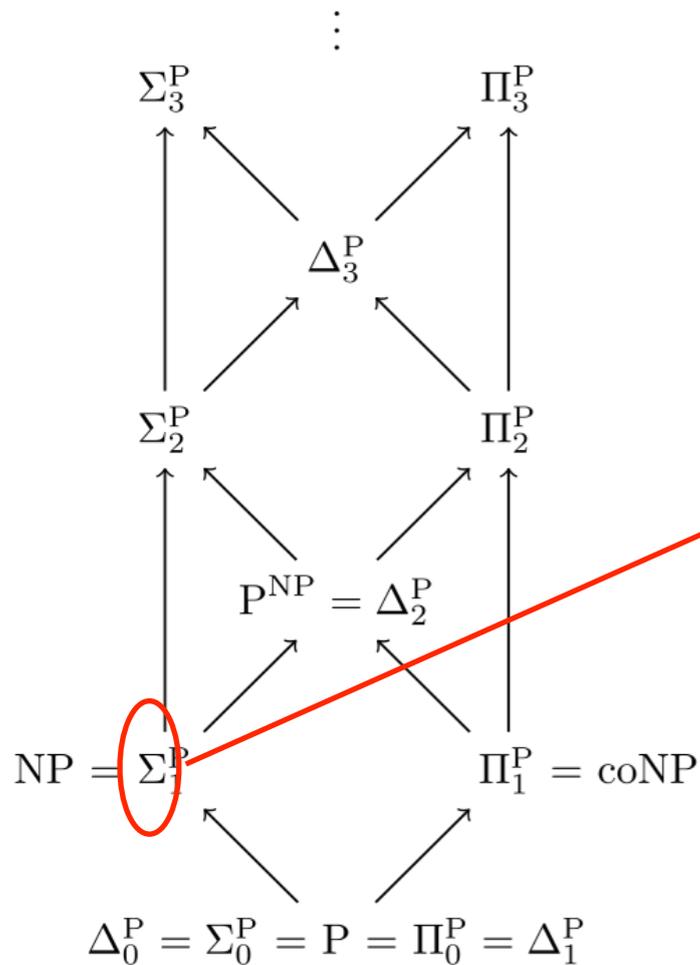
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(via formal logic, directly; a start)



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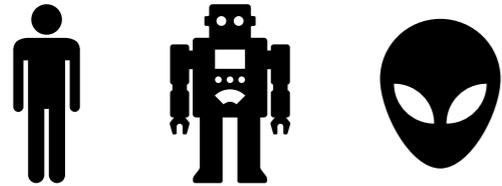
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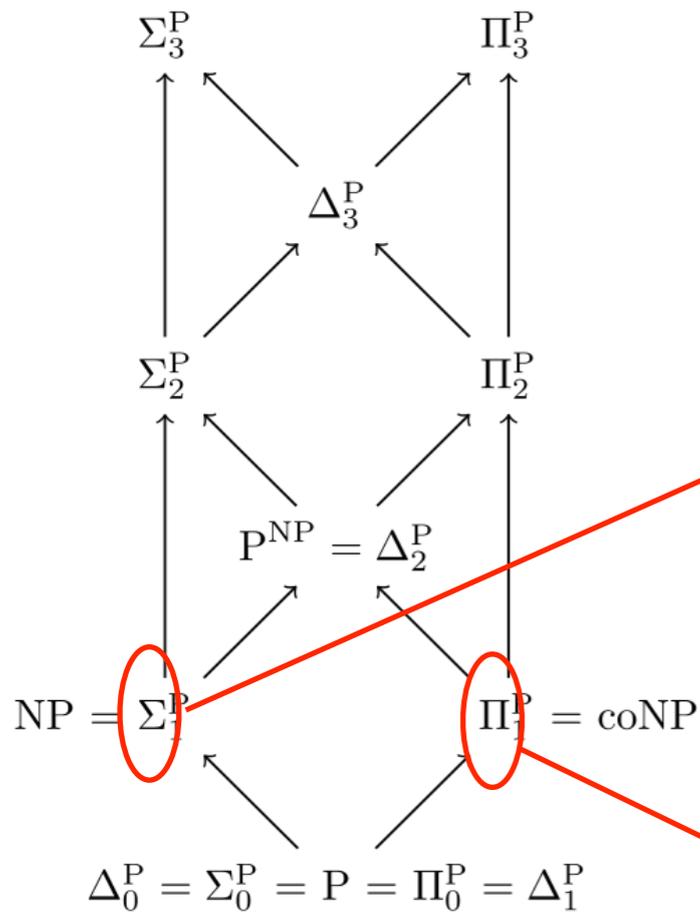
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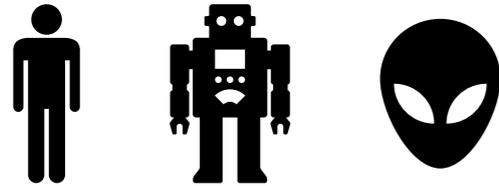
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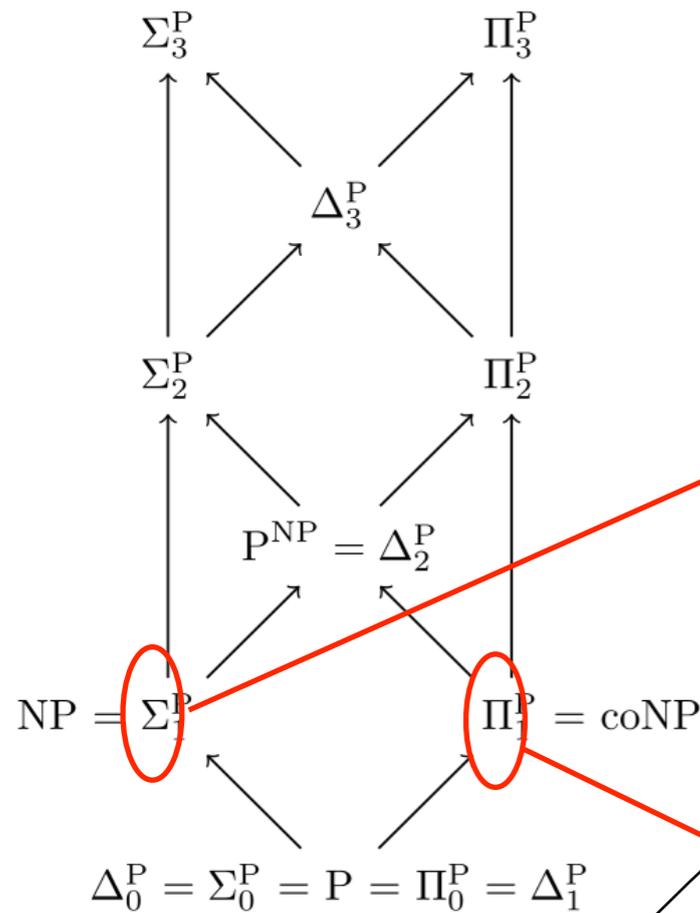
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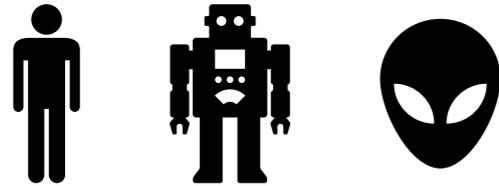
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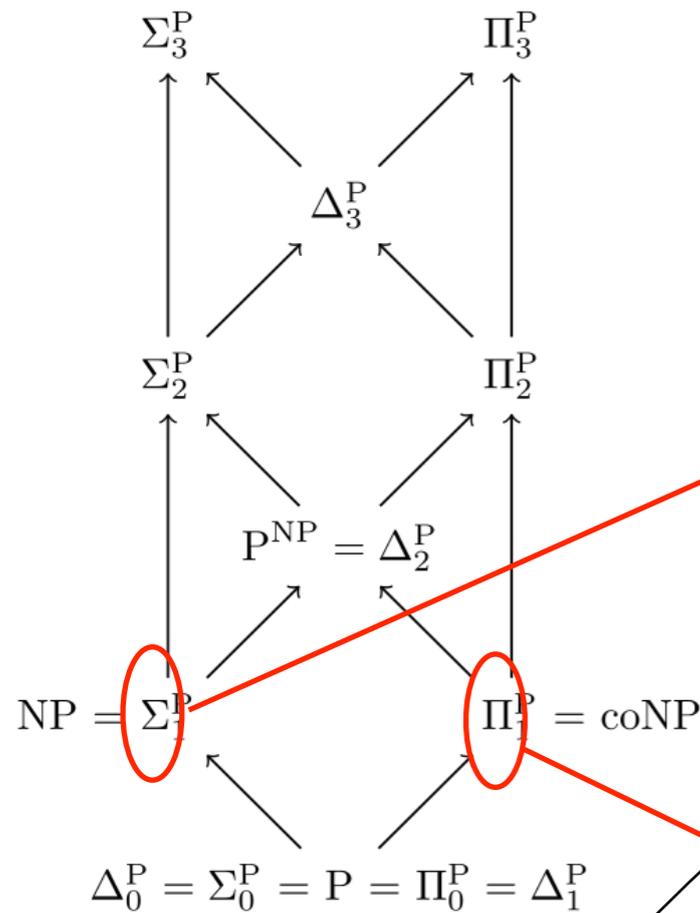
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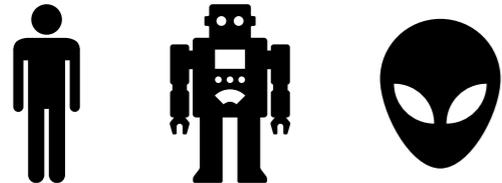
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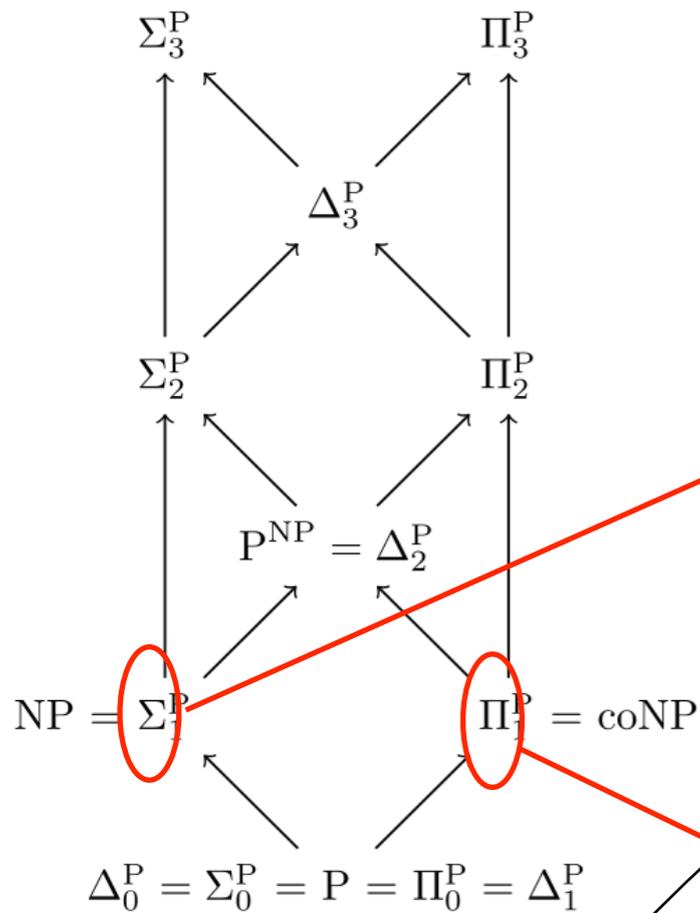
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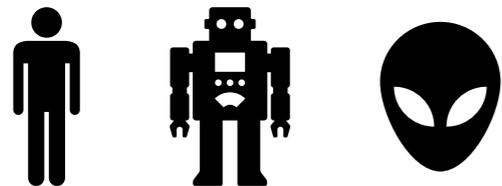
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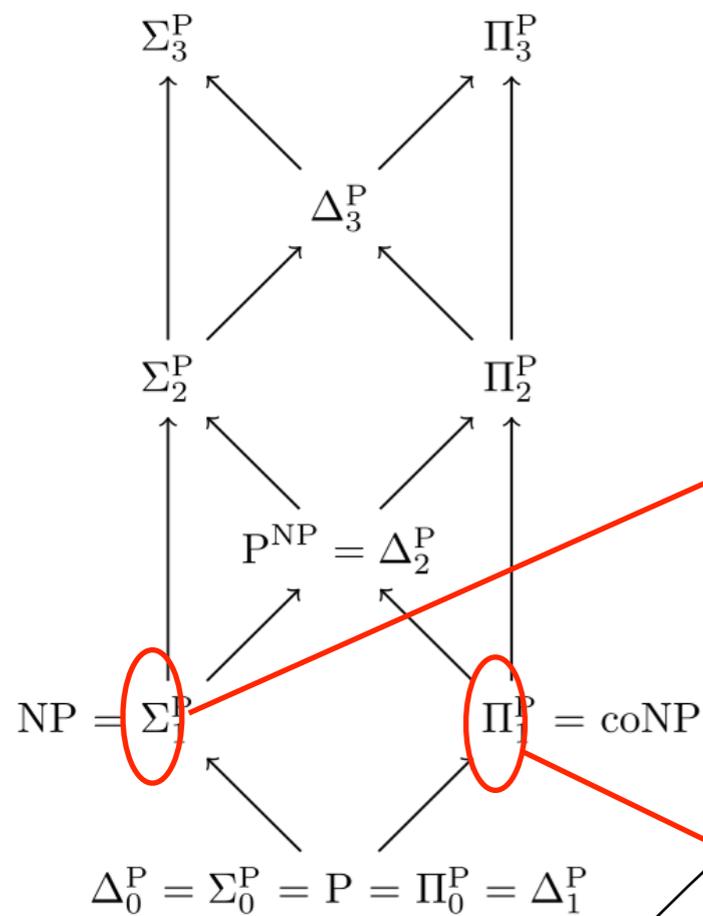
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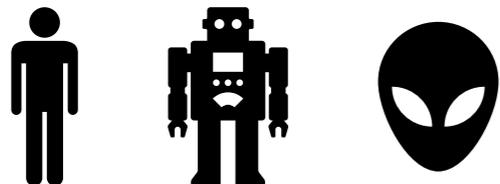
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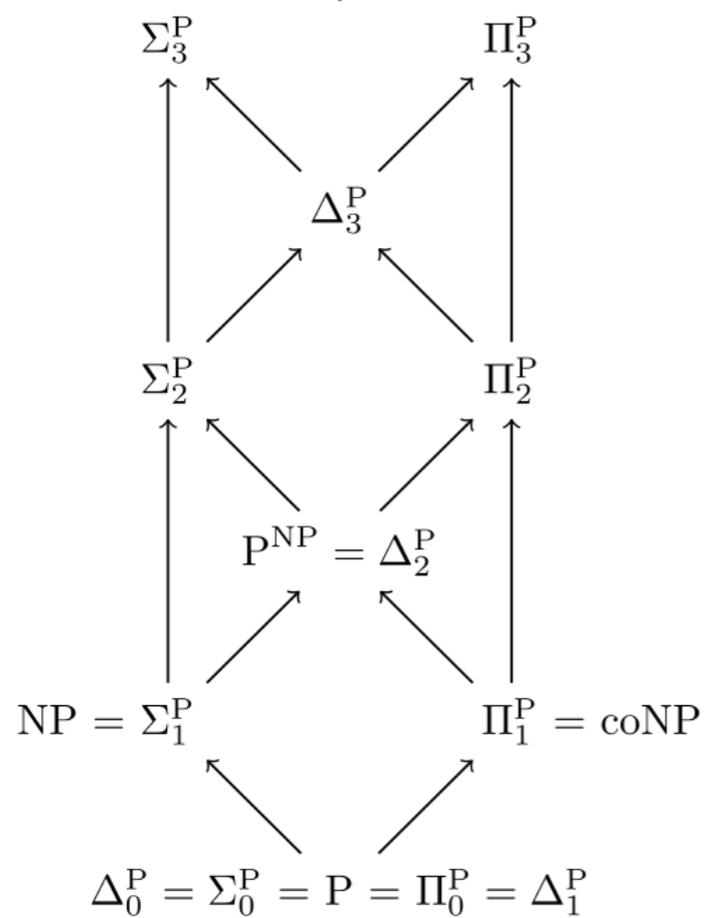
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(via formal logic, directly; a start)

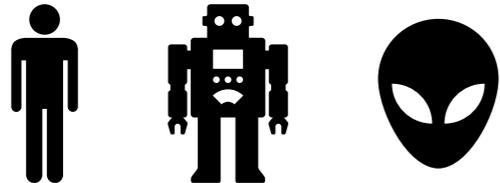


⋮

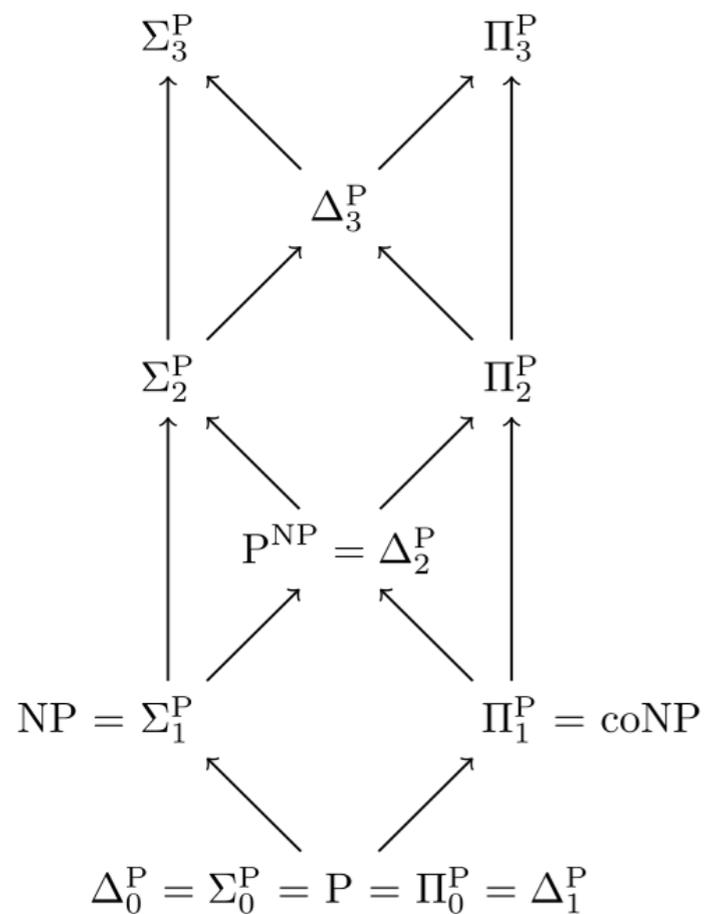


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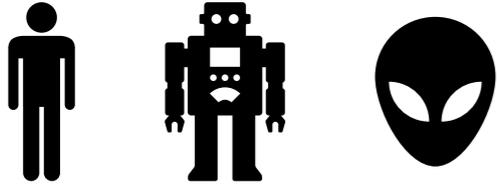
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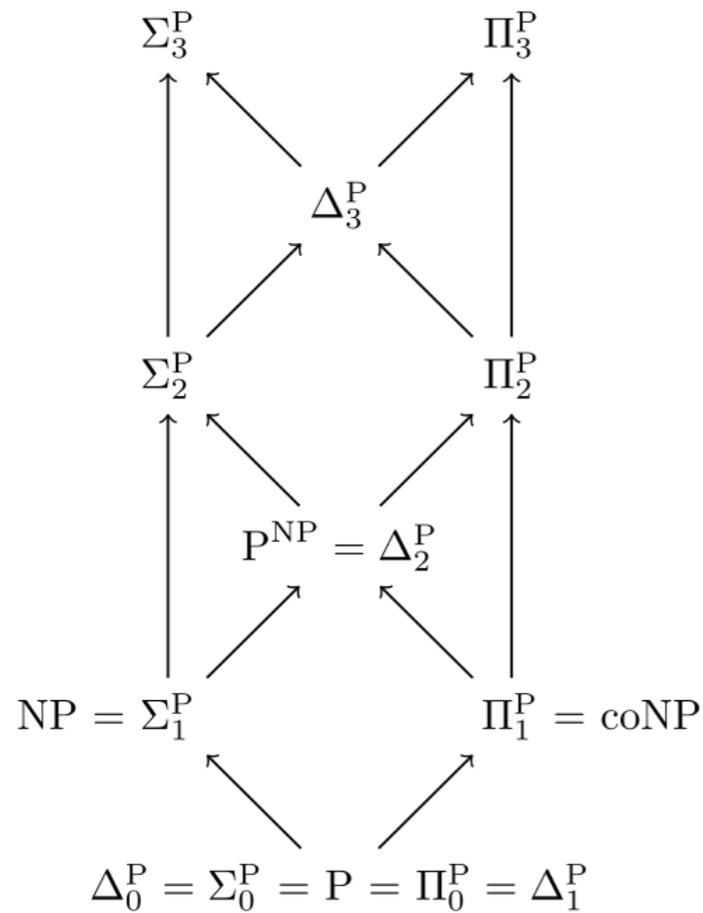
“What’s that  $\Delta$ ??”

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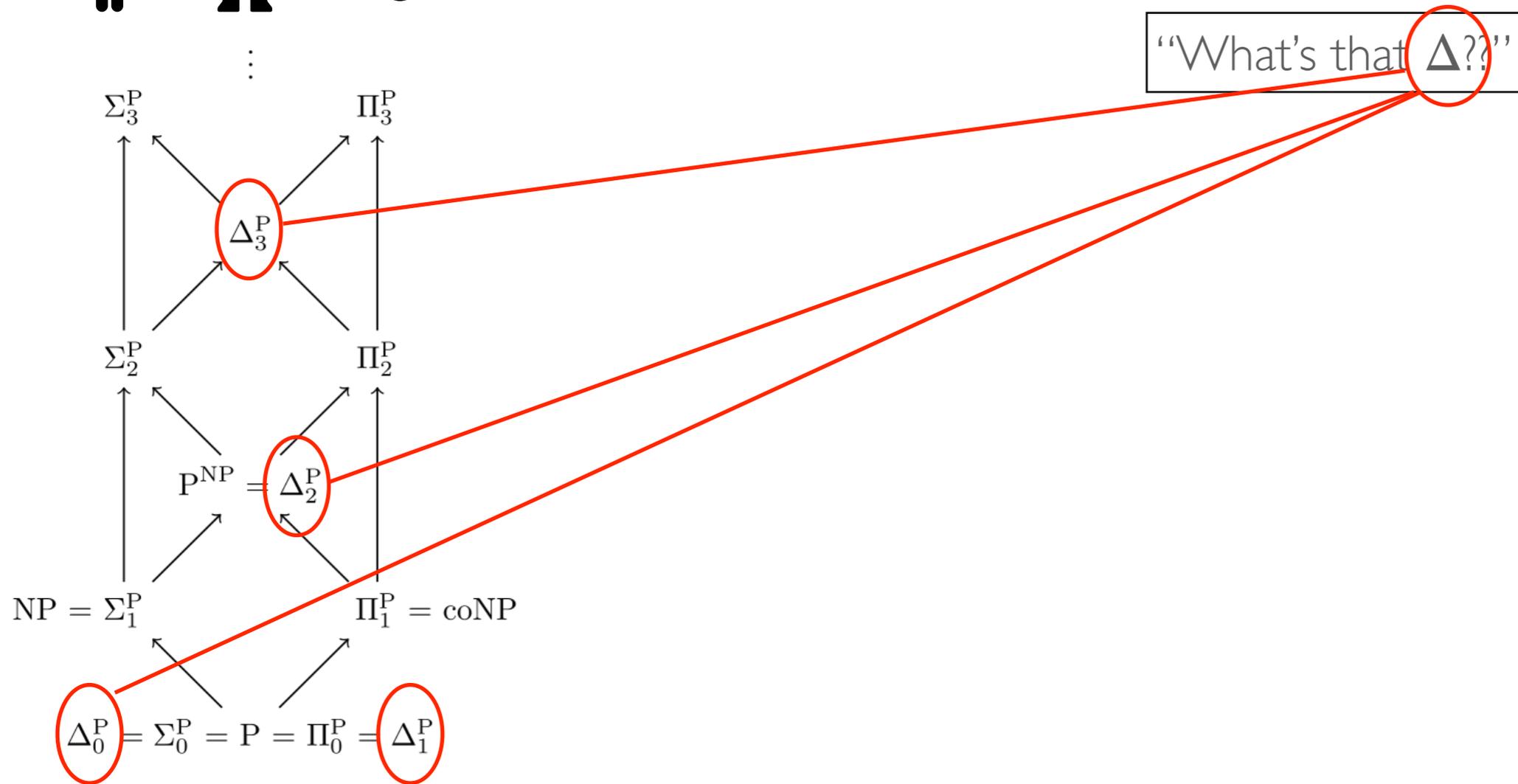
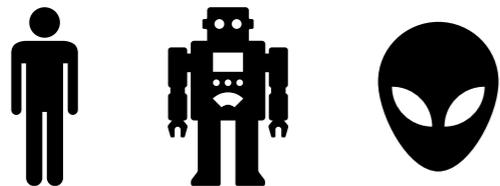
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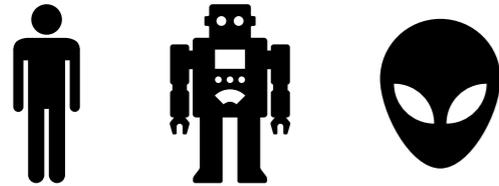
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(via formal logic, directly; a start)

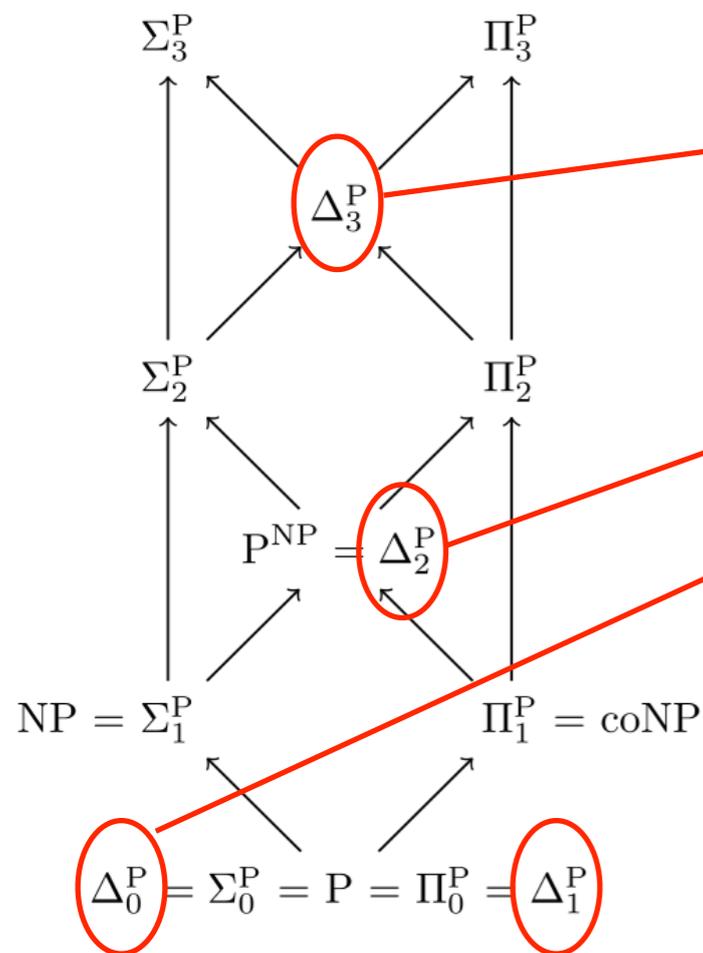


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(via formal logic, directly; a start)



⋮

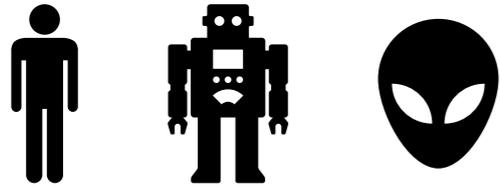


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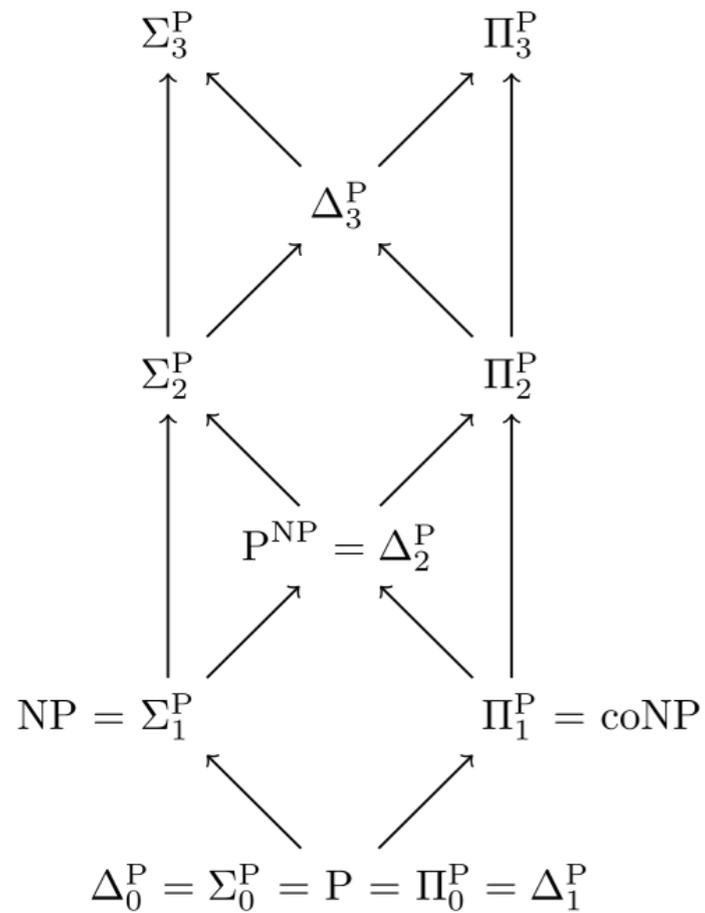
Ah. This is *not* a direct analogue to the AH. The arrows going up do indicate containment, but the purely “logician” notation based on quantifiers is apparently mixed here (dangerously). The “Delta notation” is the oracle approach to building up PH. The availability of an oracle e.g. for NP questions from P-solving machine would subsume both NP and coNP, etc.

# Polynomial Hierarchy, Part II

(via formal logic, directly)

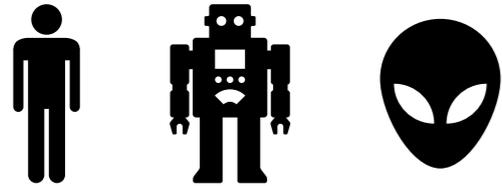


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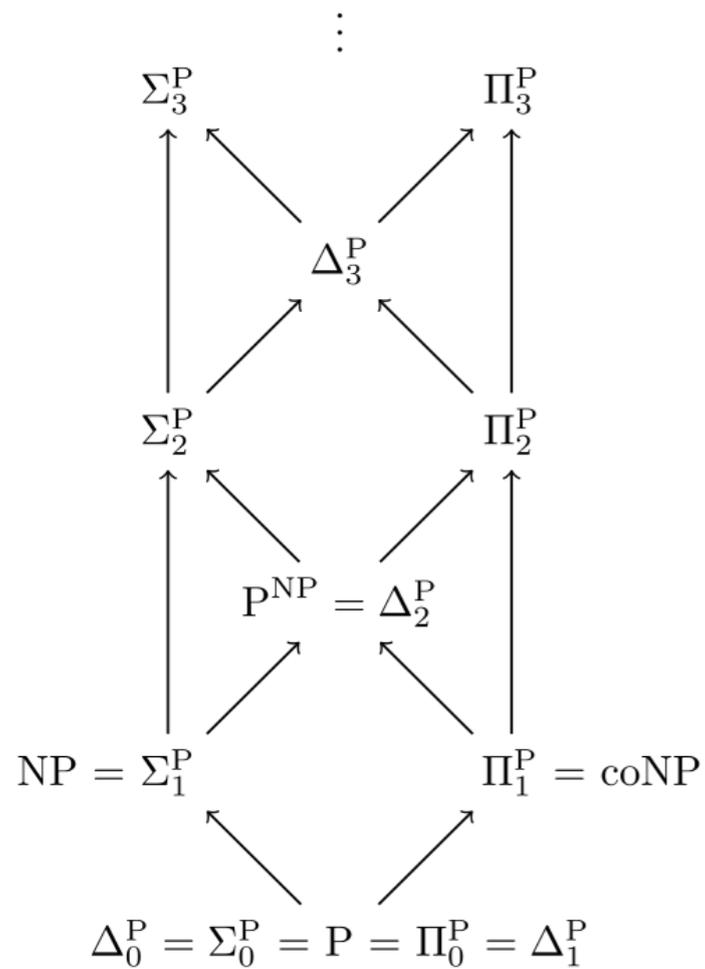


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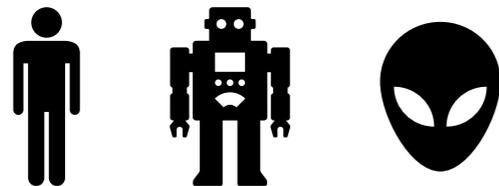


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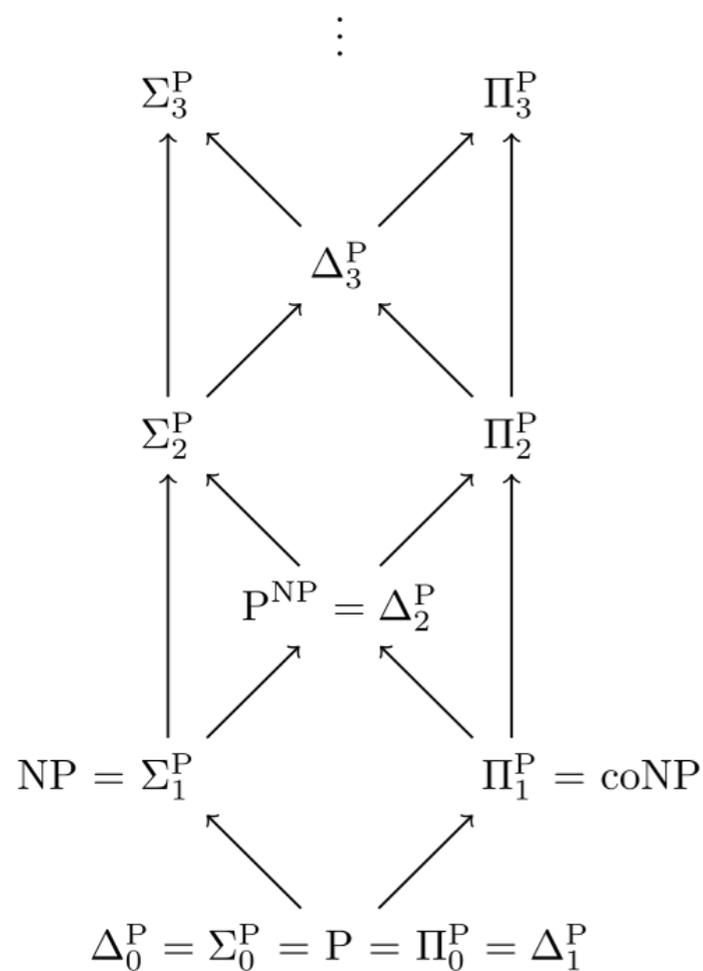
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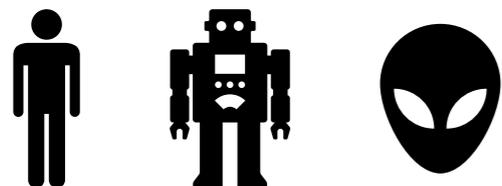
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$$\langle \phi_1, k \rangle \in L \text{ iff } \exists \phi_2 \forall \alpha KLogEquiv(\phi_1, \phi_2, |\phi_2| \leq k, \alpha(\phi_1) = \alpha(\phi_2))$$



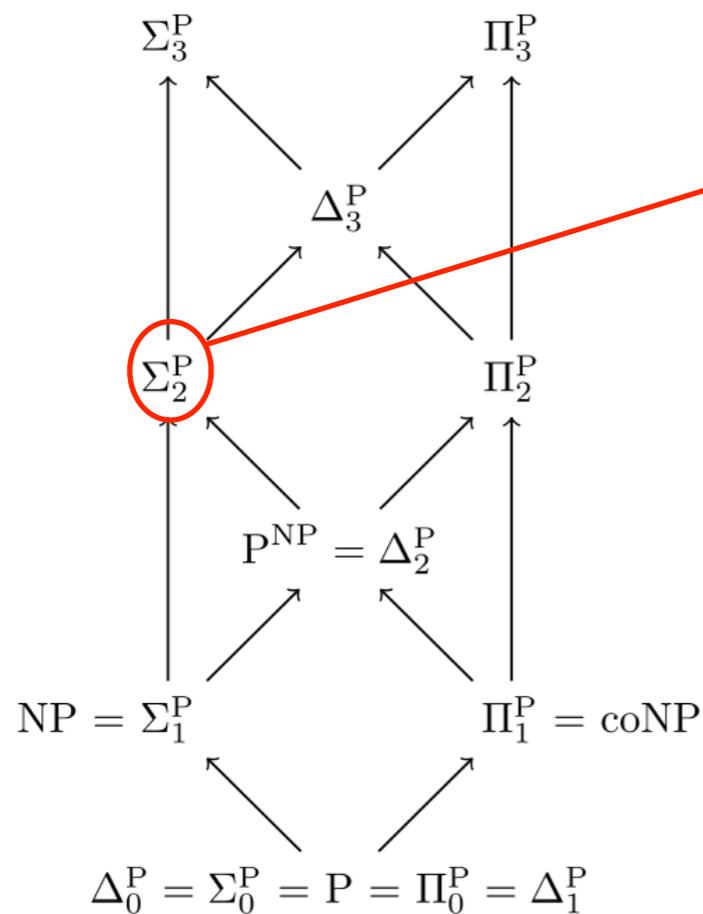
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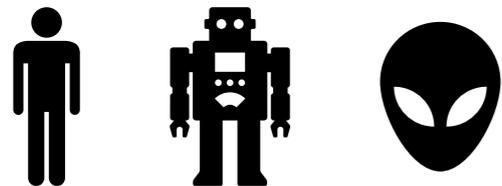
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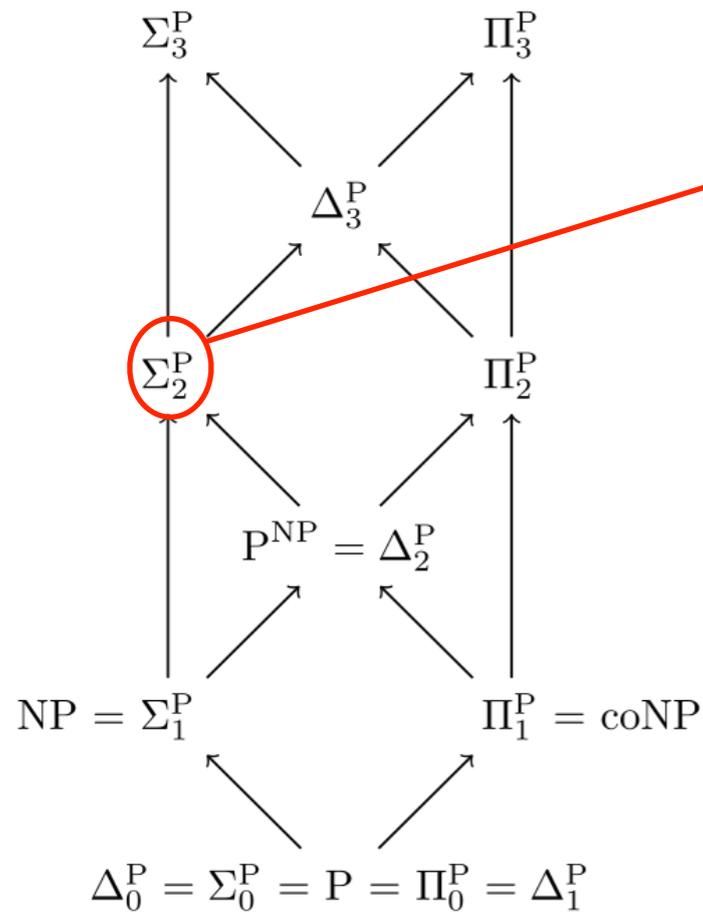


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⋮



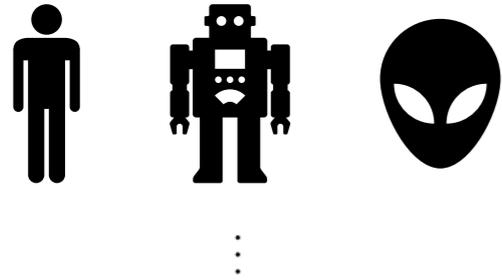
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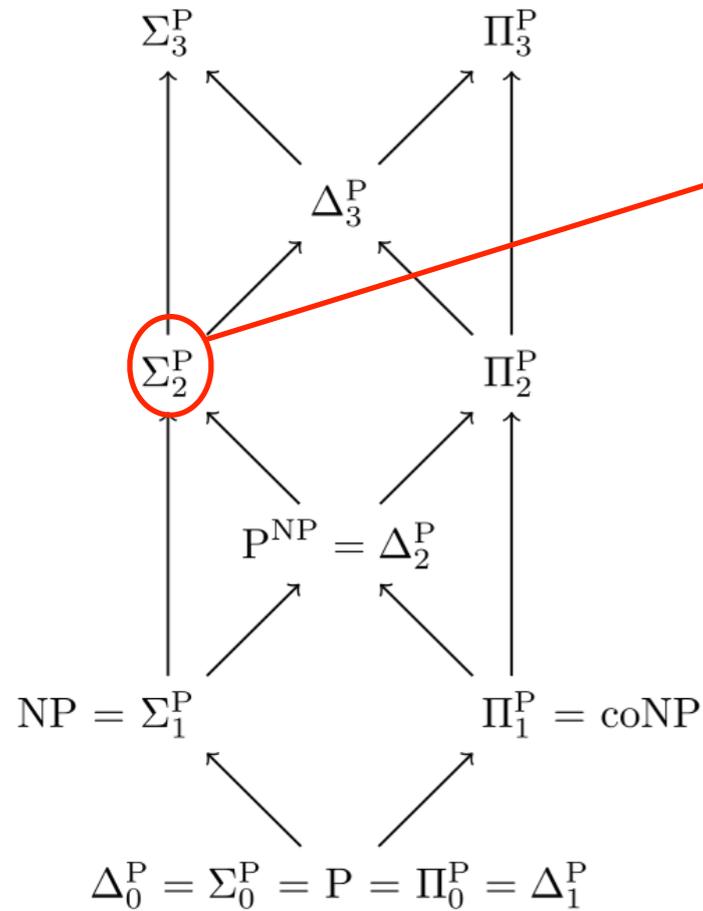
(via formal logic, directly)



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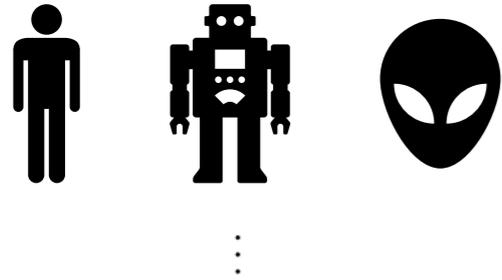
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Now we generalize:

# Polynomial Hierarchy, Part II

(via formal logic, directly)



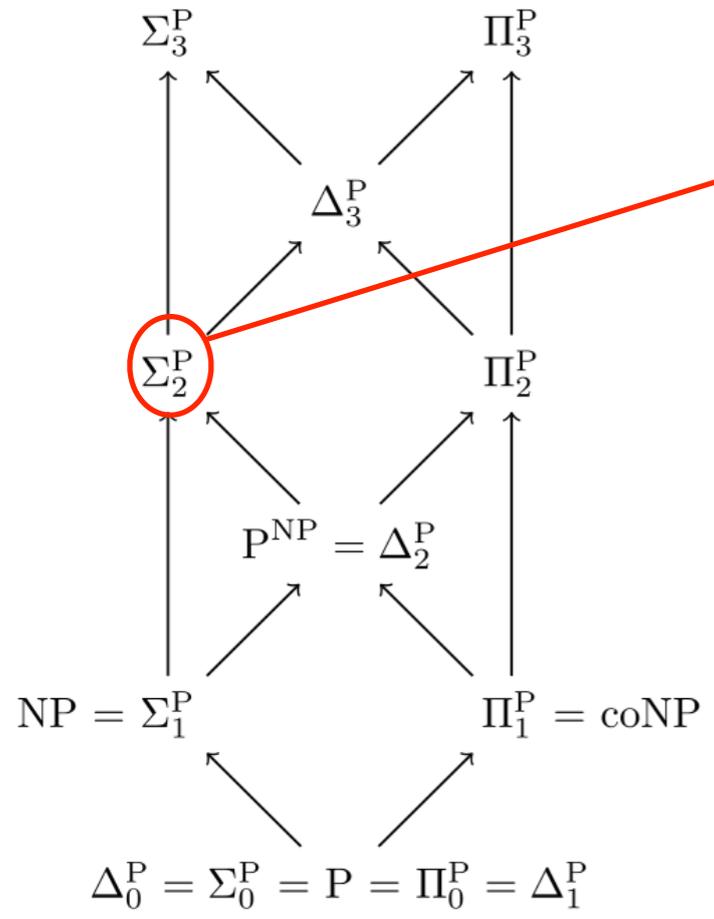
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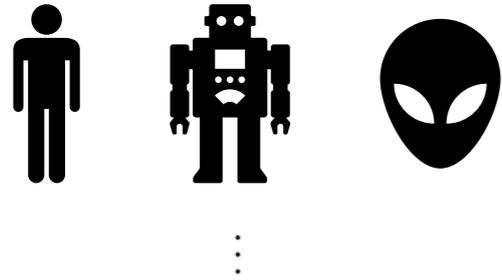
Now we generalize:

$x \in \Sigma_i \text{ iff } \exists R \exists y_1 \forall y_2 \cdots Q_i y_i R(x, y_1, y_2, \dots, y_i)$   
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# Polynomial Hierarchy, Part II

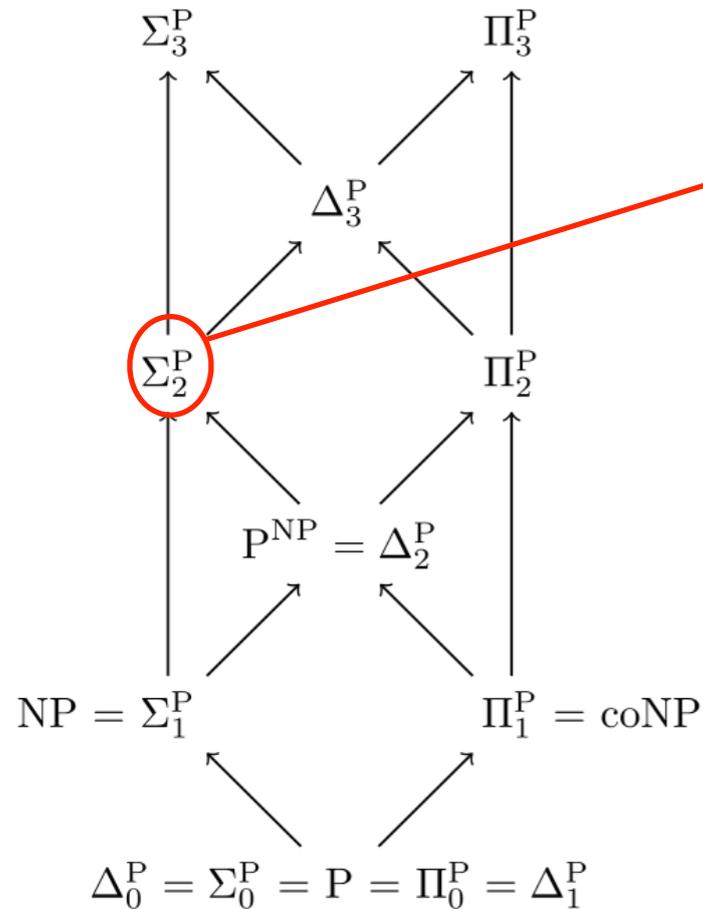
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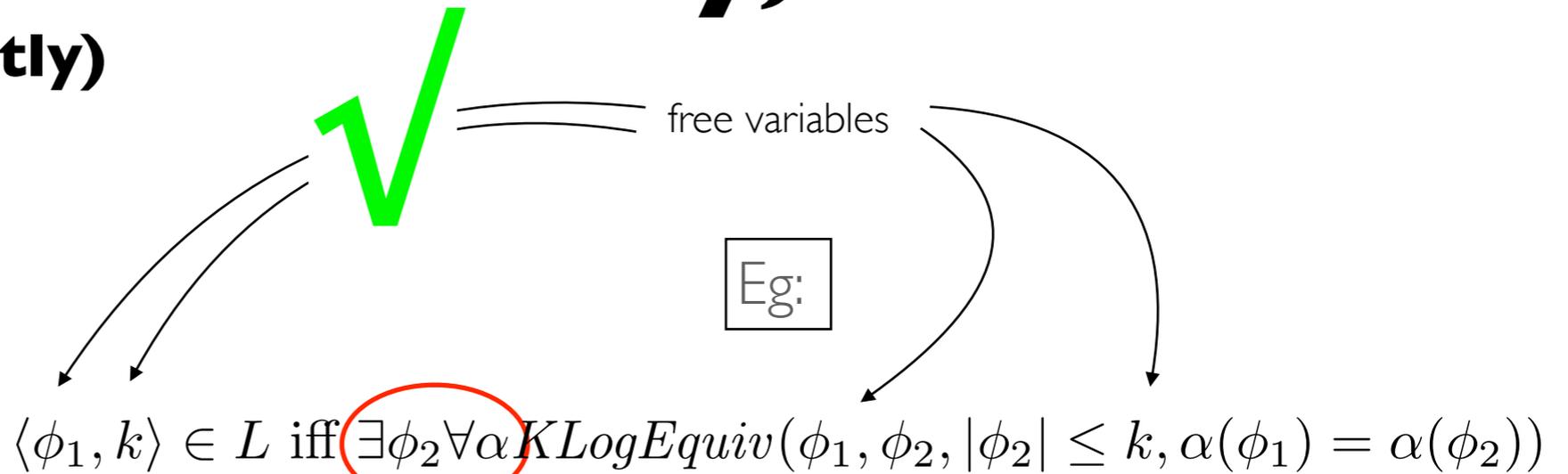
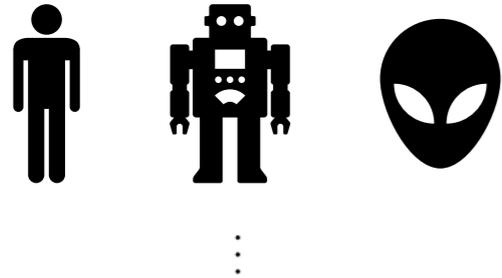
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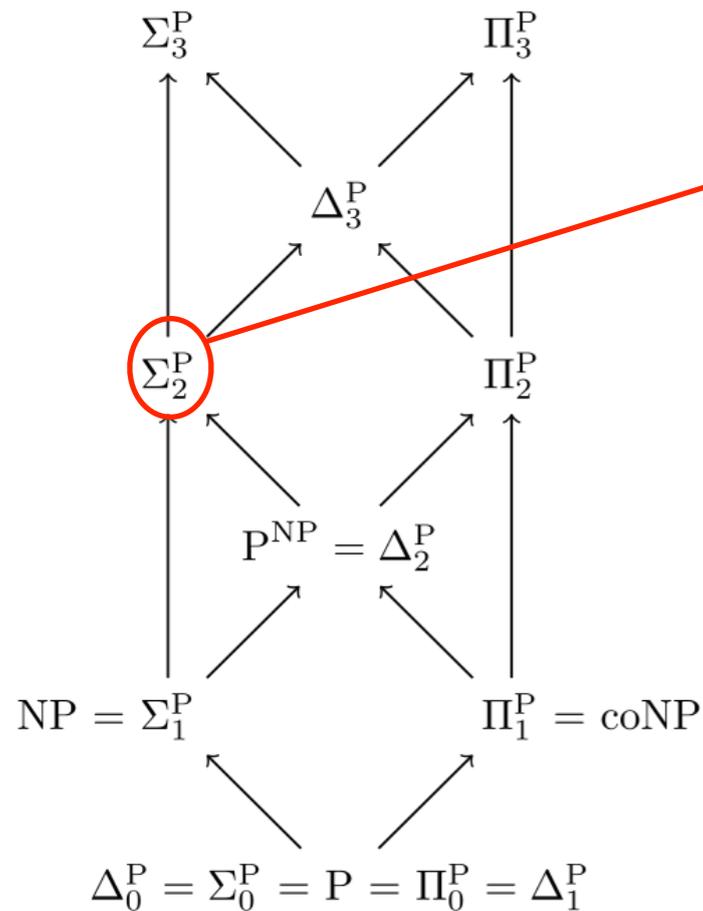
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(via formal logic, directly)



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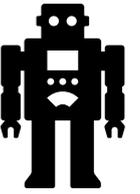
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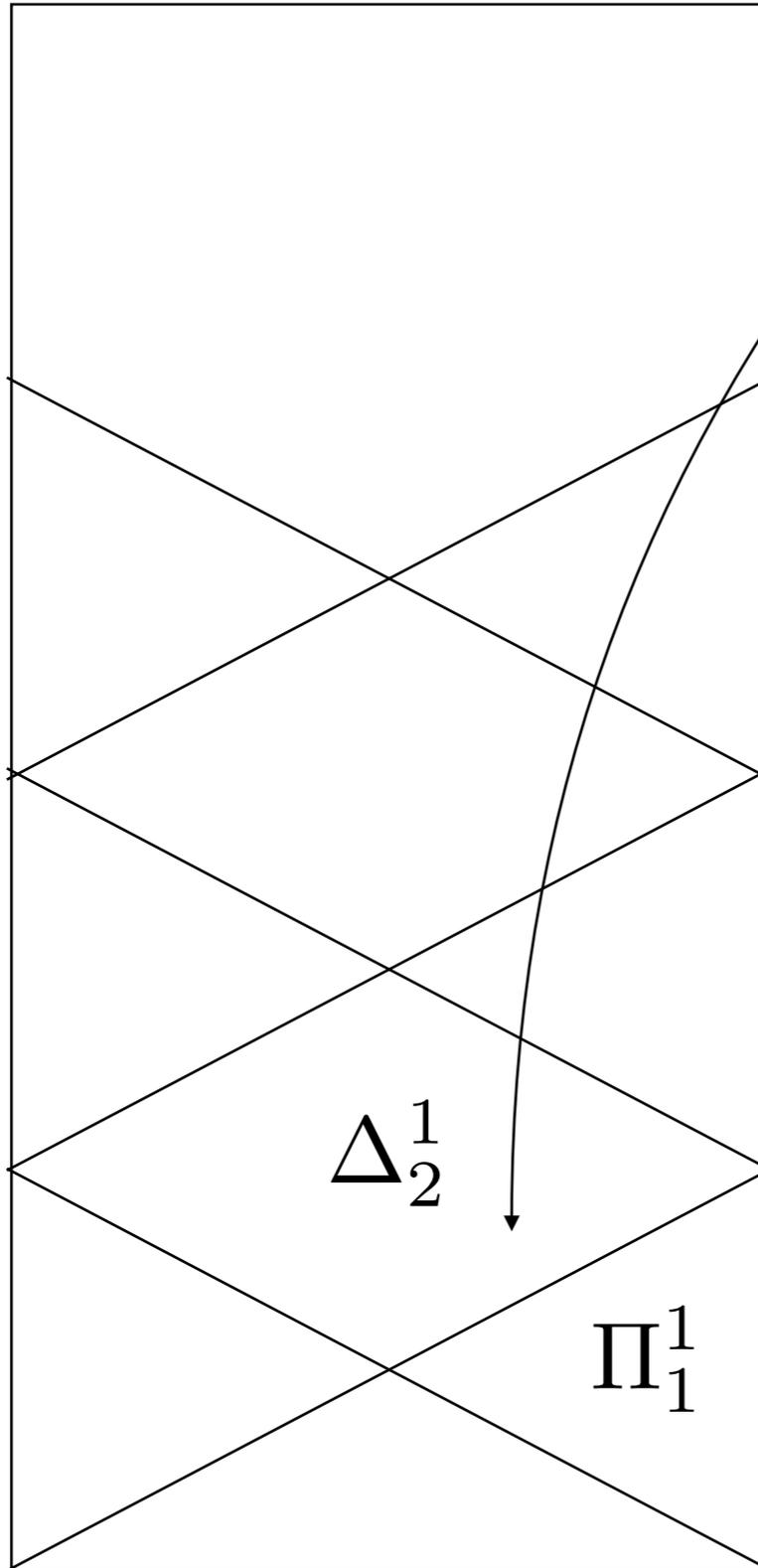
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CogSci and AI need to say more about where AI falls/can fall in the landscape.

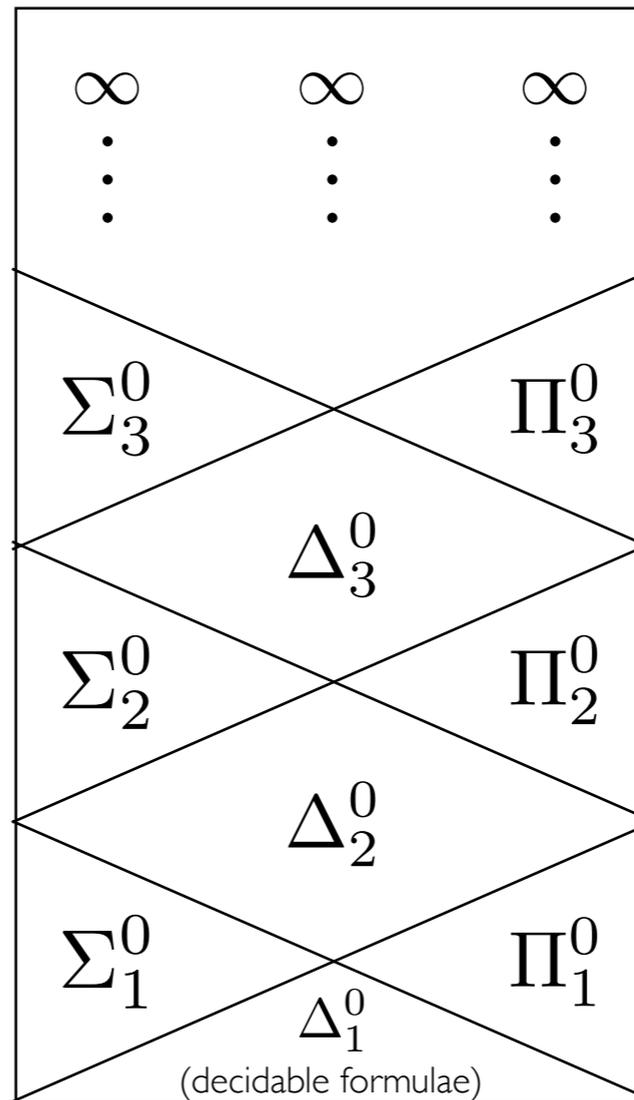


$A^n \mathcal{H}$  (Analytic Hierarchy)



Infinite Time Turing Machines (ITTTMs)

$A^r \mathcal{H}$  (Arithmetic Hierarchy)

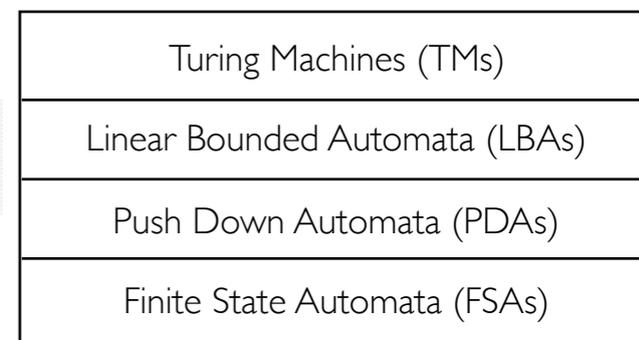


Human Persons (according to Bringsjord)

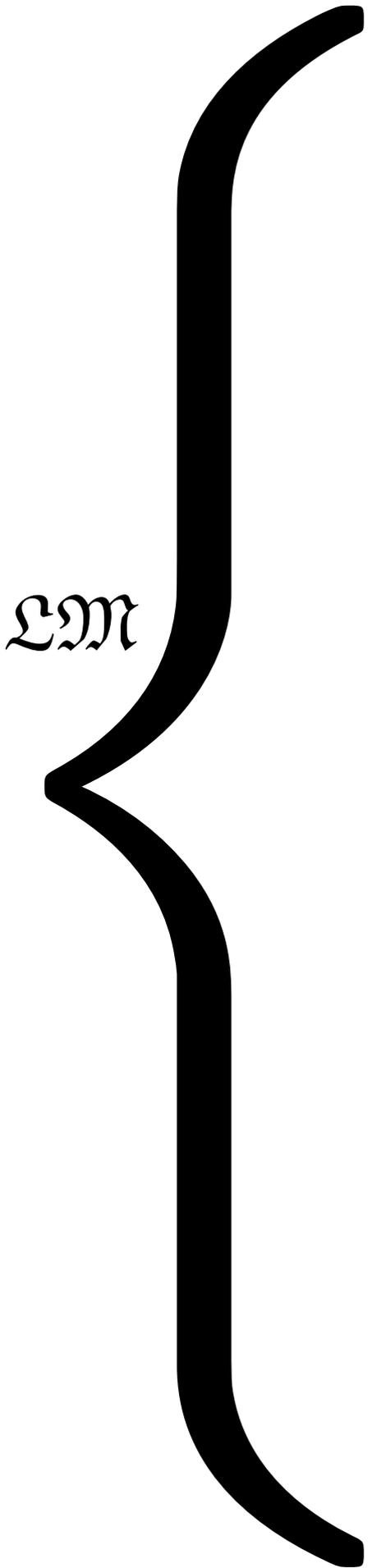
Human Brains (according to Granger)



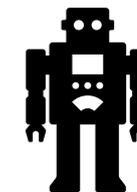
$\mathcal{CH}$  (Chomsky Hierarchy)



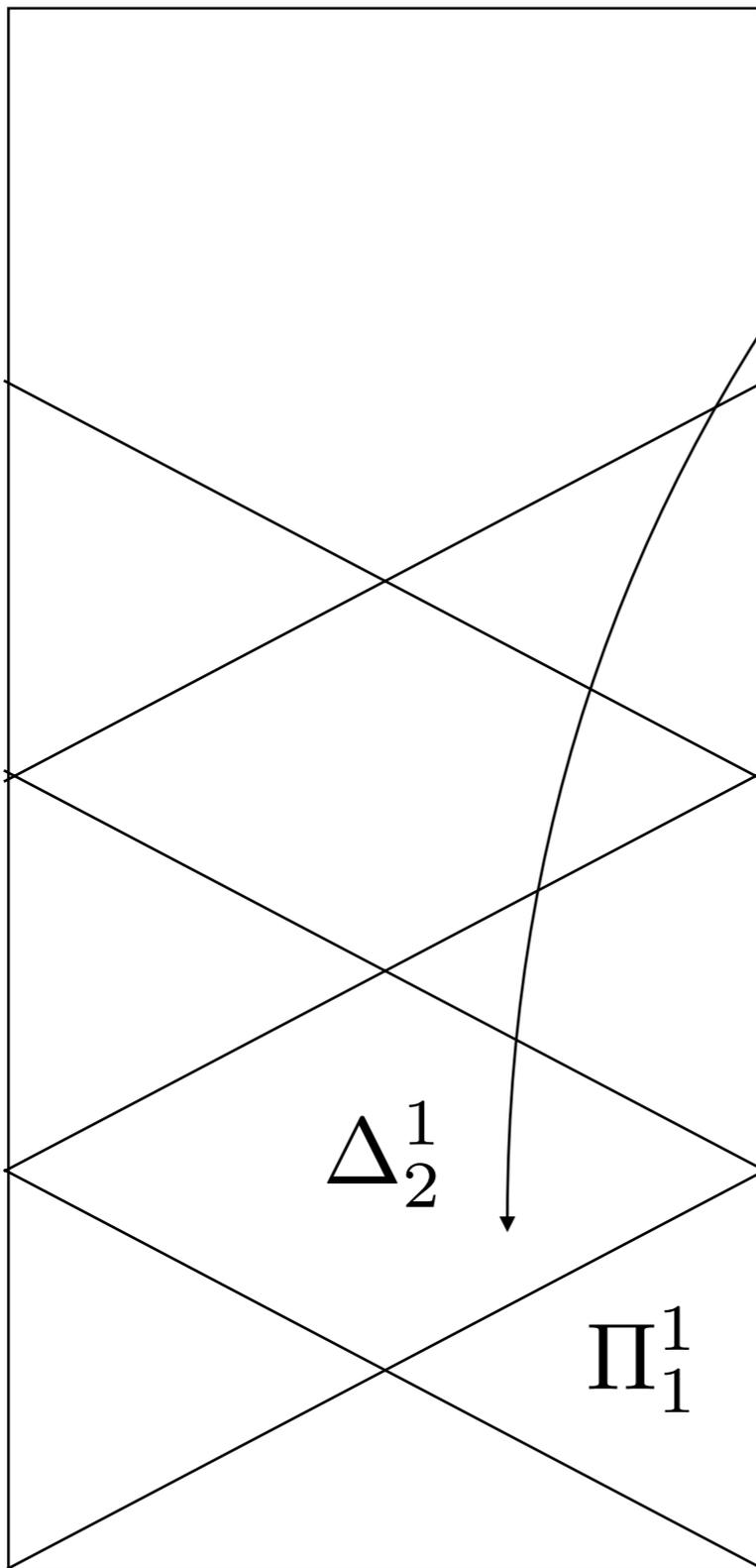
$\mathcal{EM}$



CogSci and AI need to say more about where AI falls/can fall in the landscape.

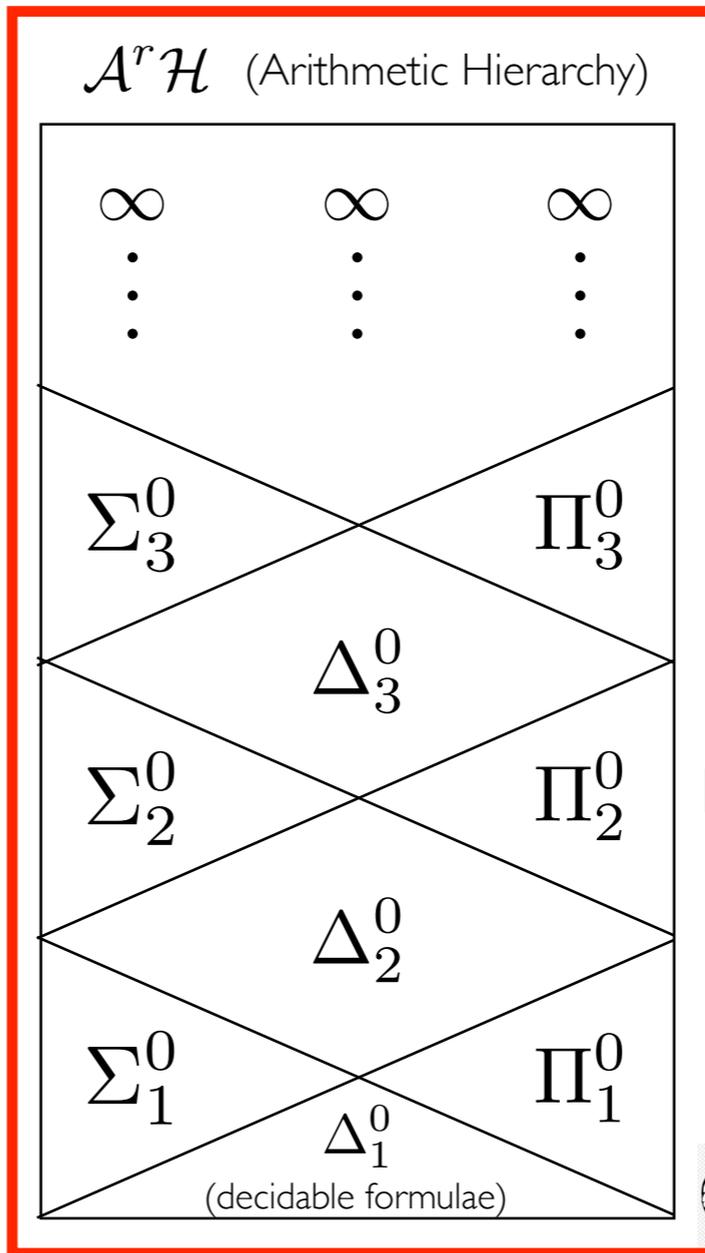


$A^n \mathcal{H}$  (Analytic Hierarchy)



Infinite Time Turing Machines (ITTMs)

$A^r \mathcal{H}$  (Arithmetic Hierarchy)



Human Persons (according to Bringsjord)

Human Brains (according to Granger)

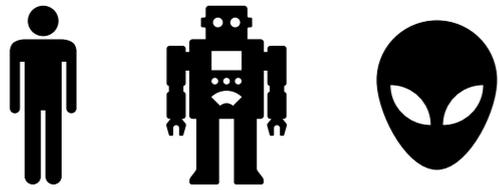


$\mathcal{CH}$  (Chomsky Hierarchy)

- Turing Machines (TMs)
- Linear Bounded Automata (LBAs)
- Push Down Automata (PDAs)
- Finite State Automata (FSAs)

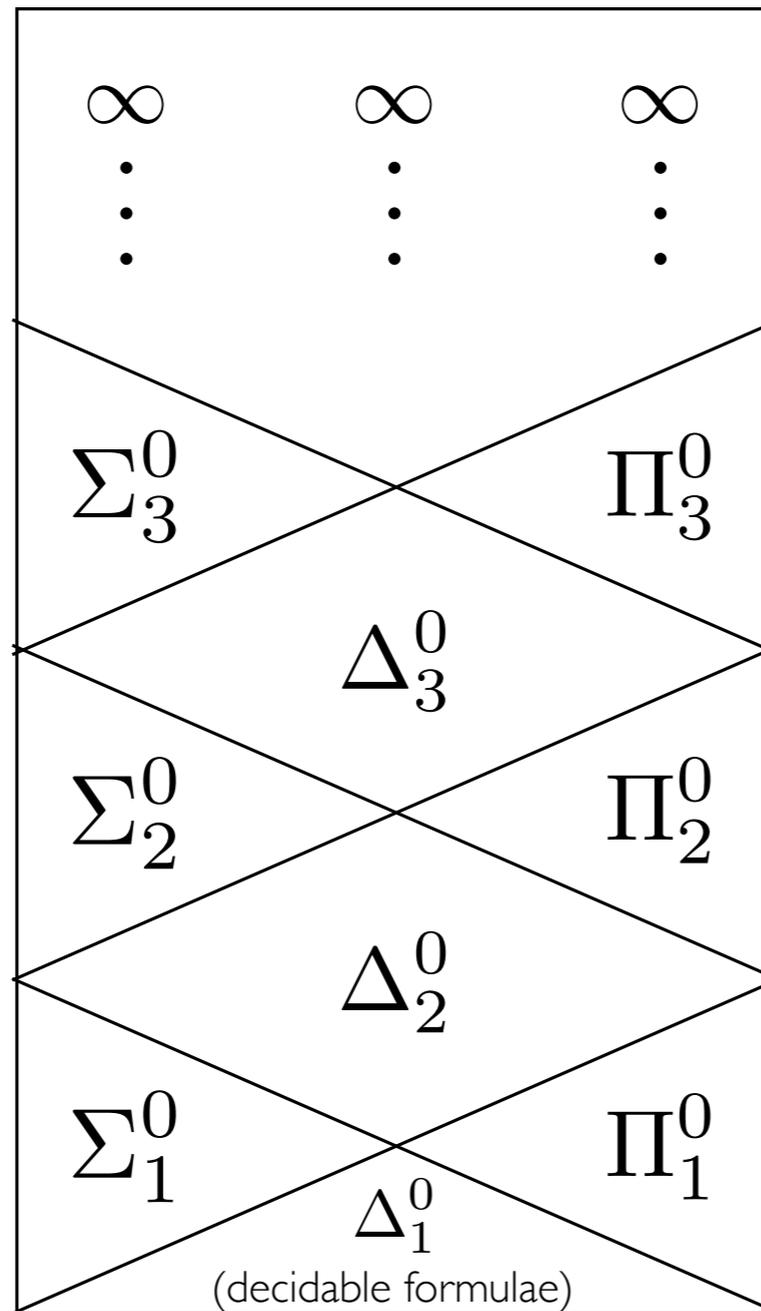


$\mathcal{EM}$



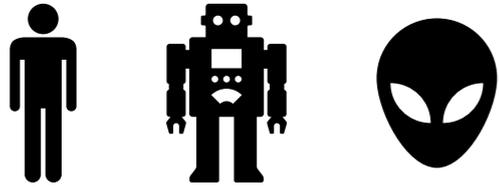
$$2\text{SAMEFUNC} := \{m_1, m_2 : \forall u \forall v [\exists k (\langle m_1, u \rangle : v, k \leftrightarrow \exists k' (\langle m_2, u \rangle : v, k'))]\}$$

$\mathcal{A}^r \mathcal{H}$  (Arithmetic Hierarchy)



$$\Delta_1^0 = \Sigma_1^0 \cap \Pi_1^0$$

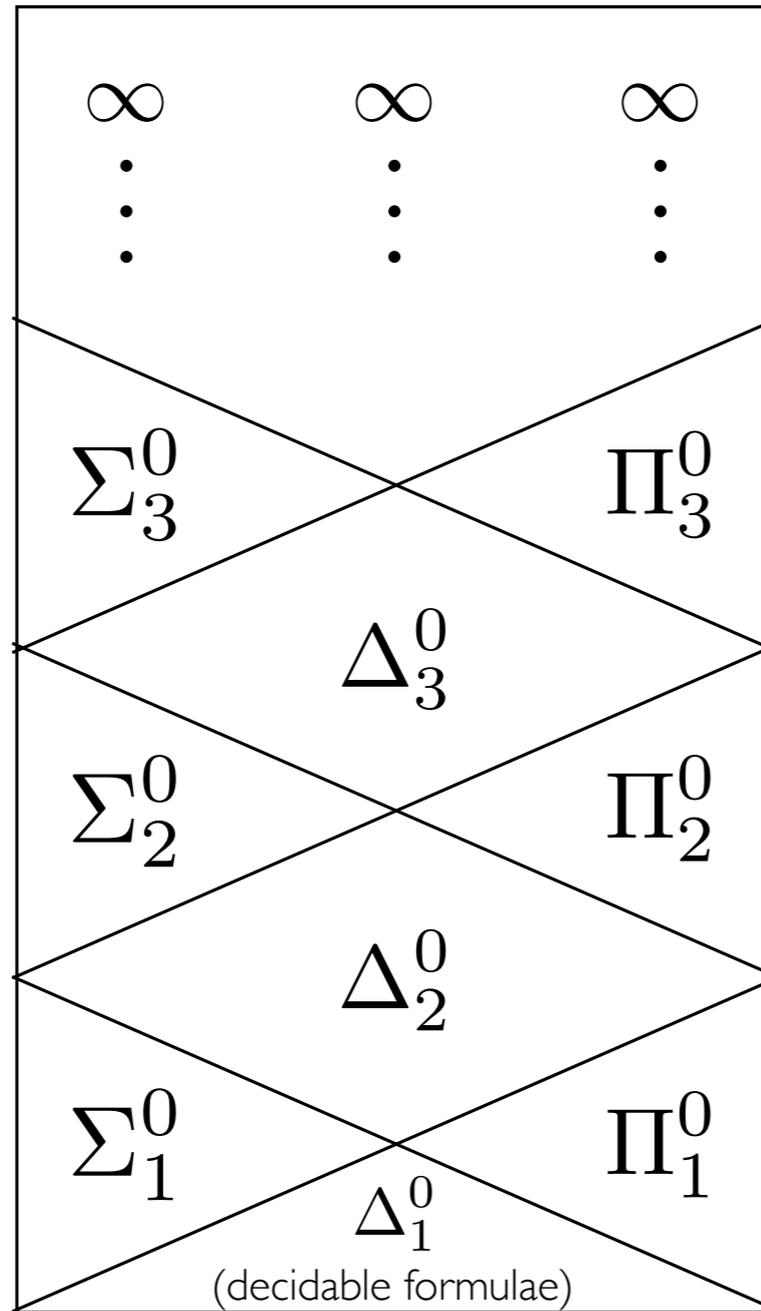
# Arithmetic Hierarchy, Part I



Can you see the carryover from PH?

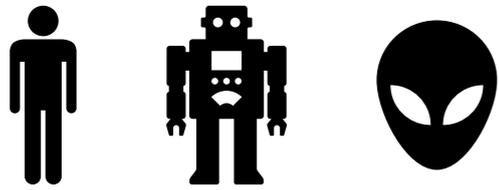
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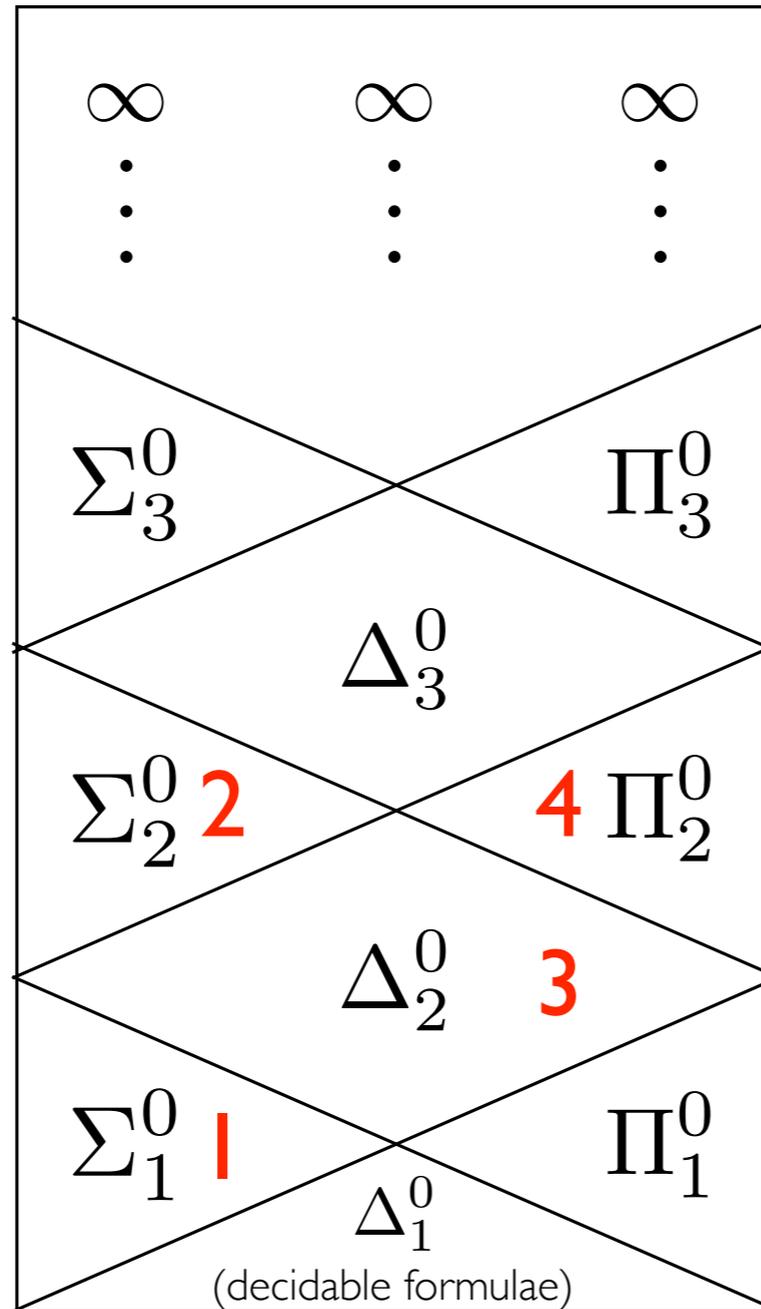
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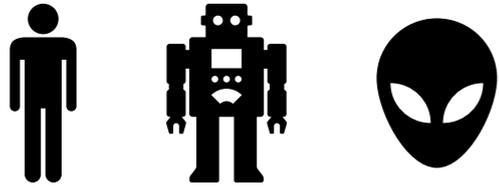
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1 2 3 or 4?

semi-decidable

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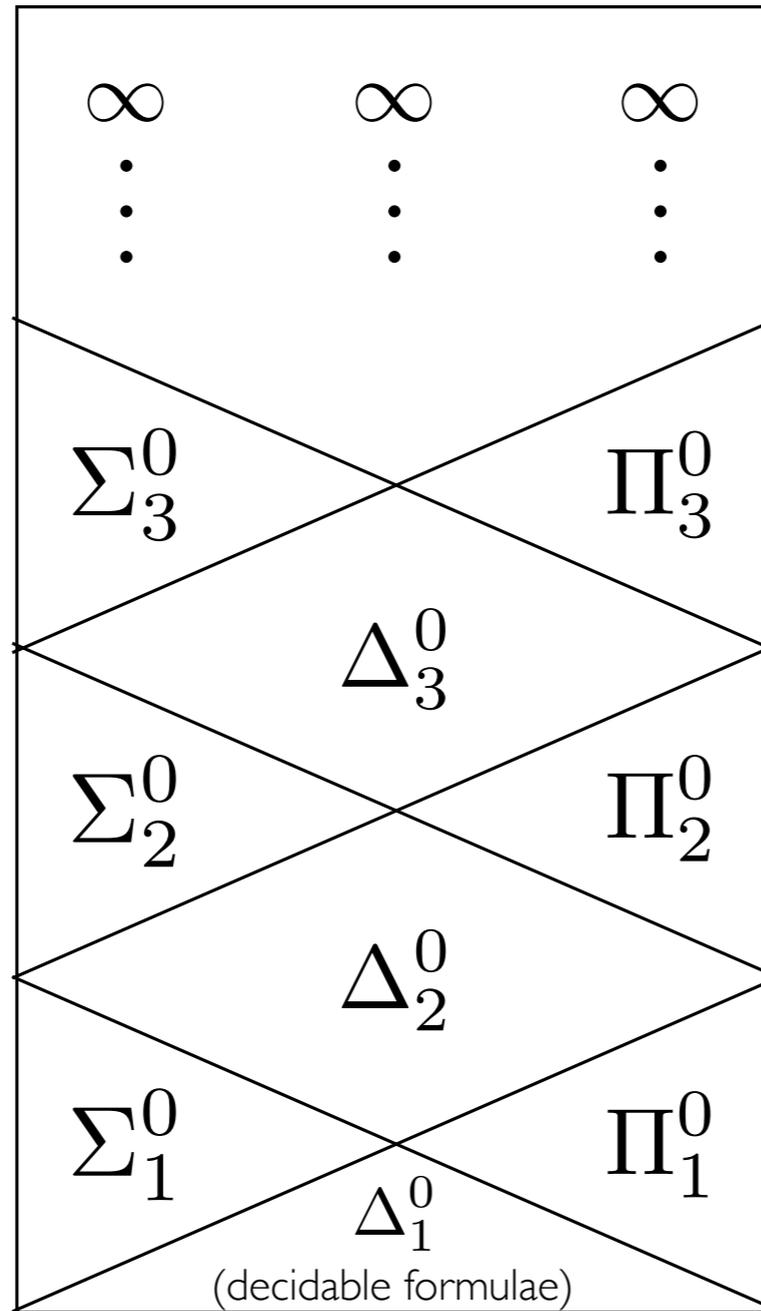
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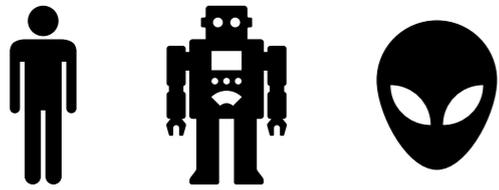
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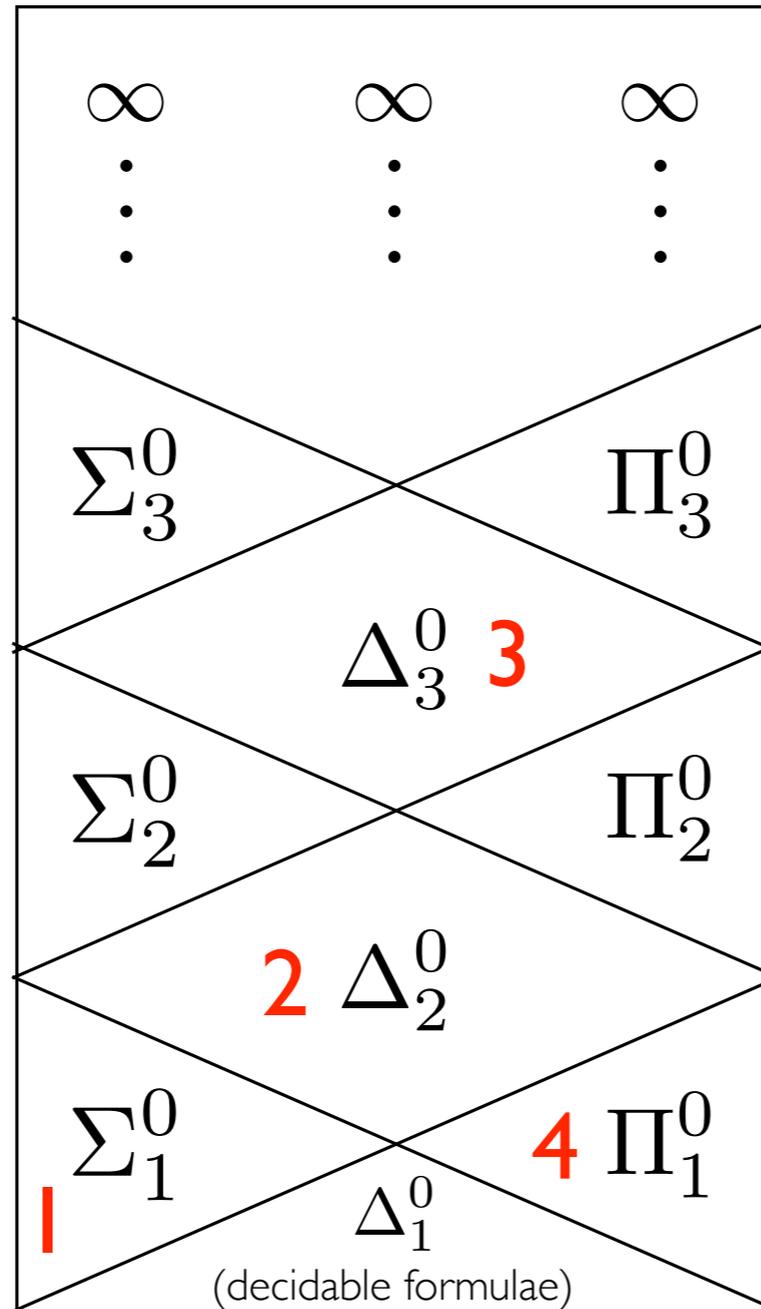
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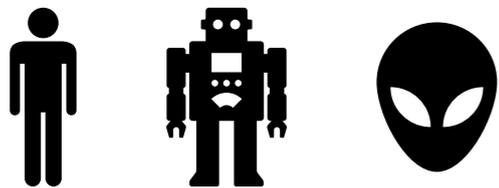
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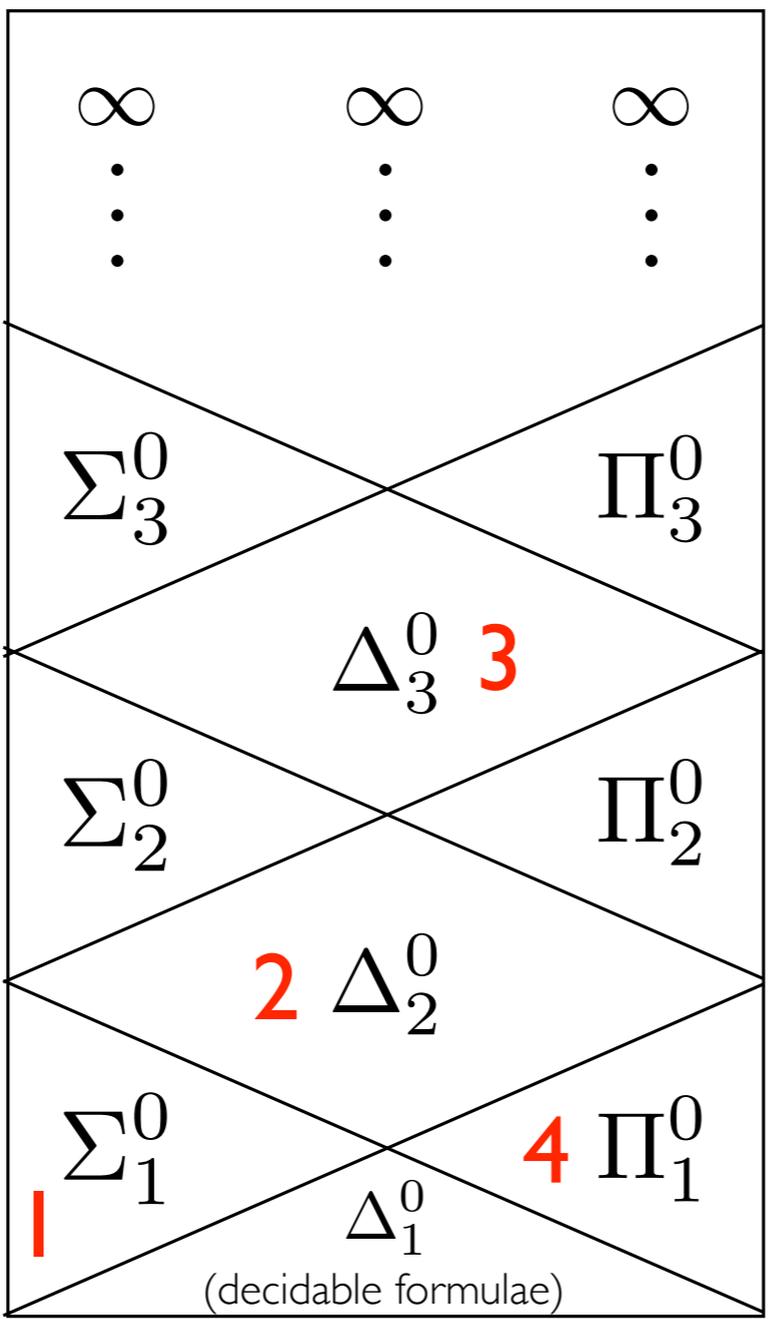
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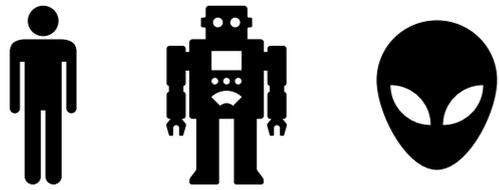
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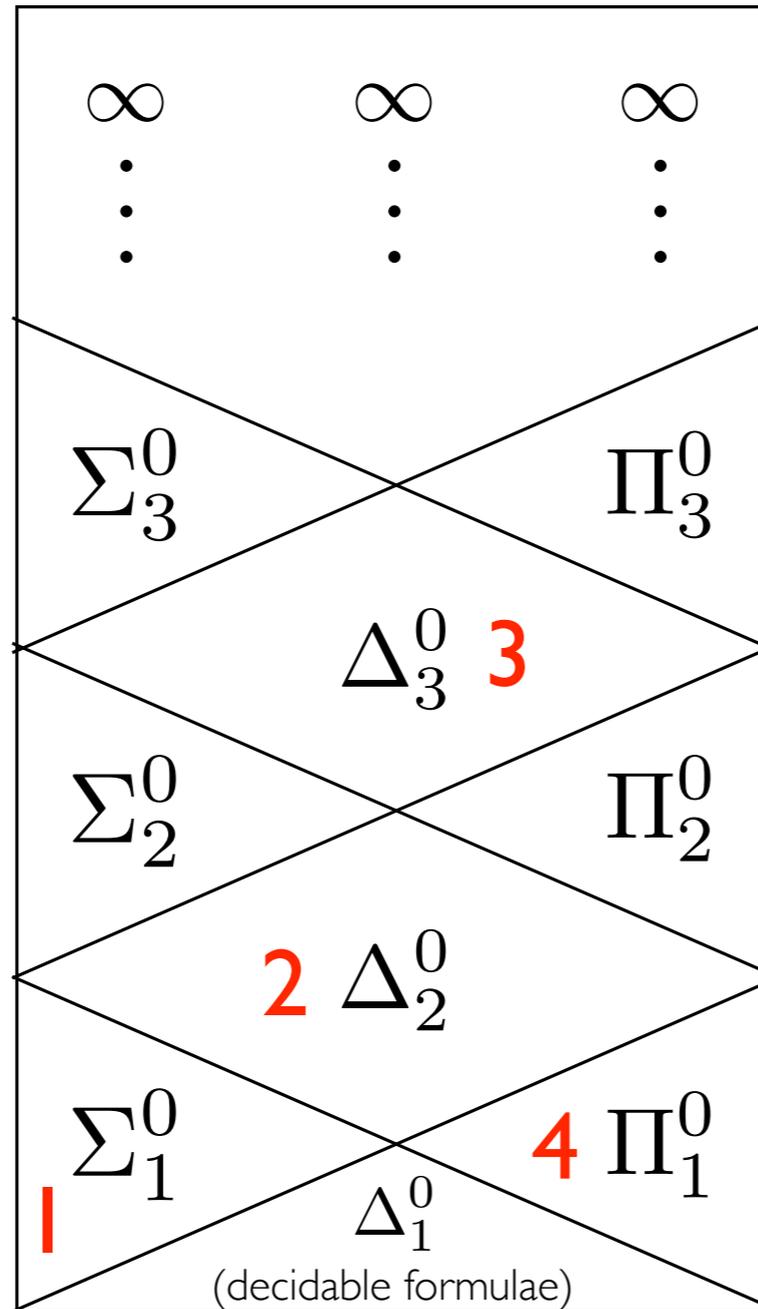
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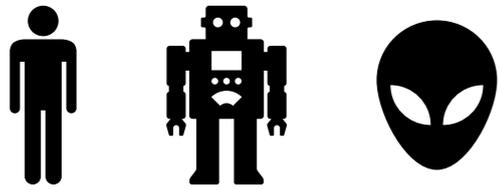
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semi-decidable



$$\Delta_1^0 = \Sigma_1^0 \cap \Pi_1^0$$

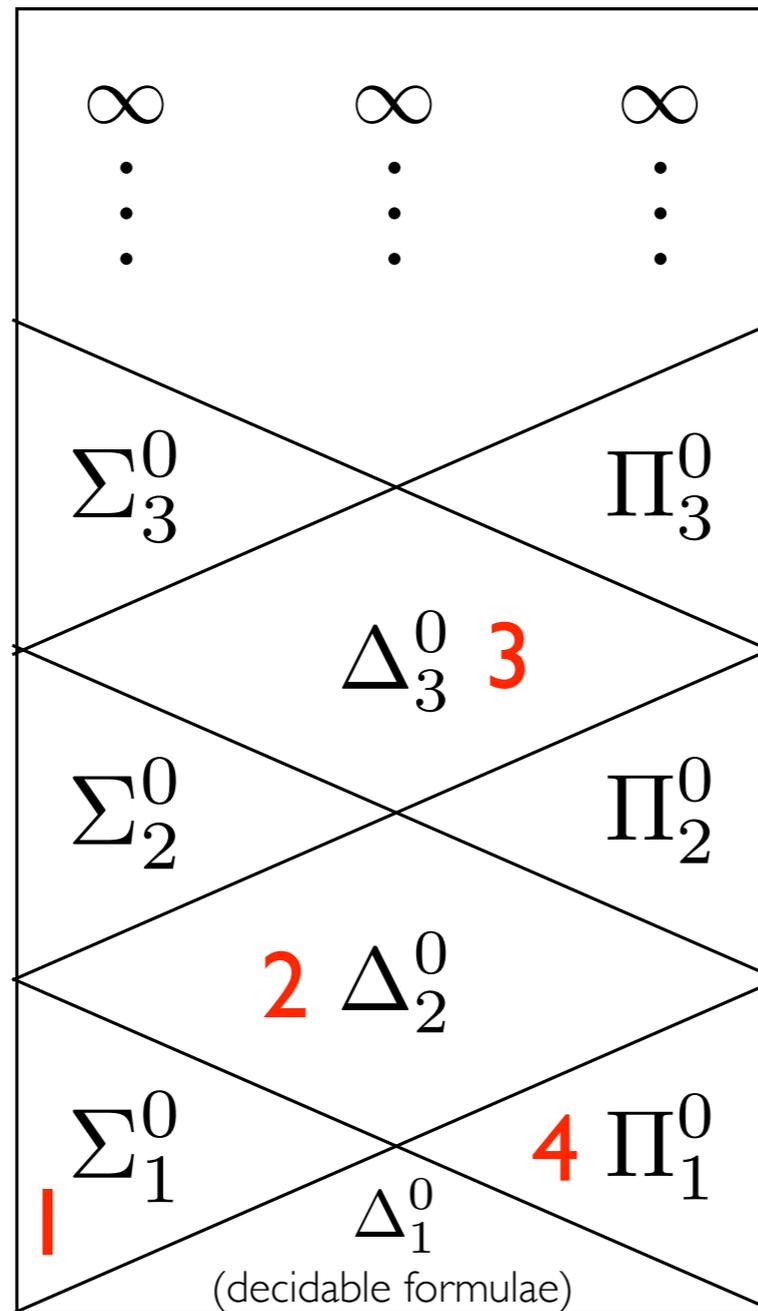
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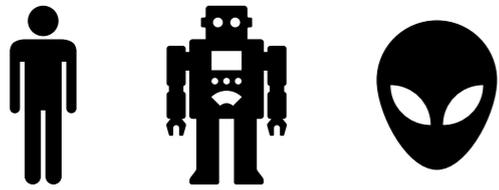
Try your hand at classifying! ...

semi-decidable



$$\Delta_1^0 = \Sigma_1^0 \cap \Pi_1^0$$

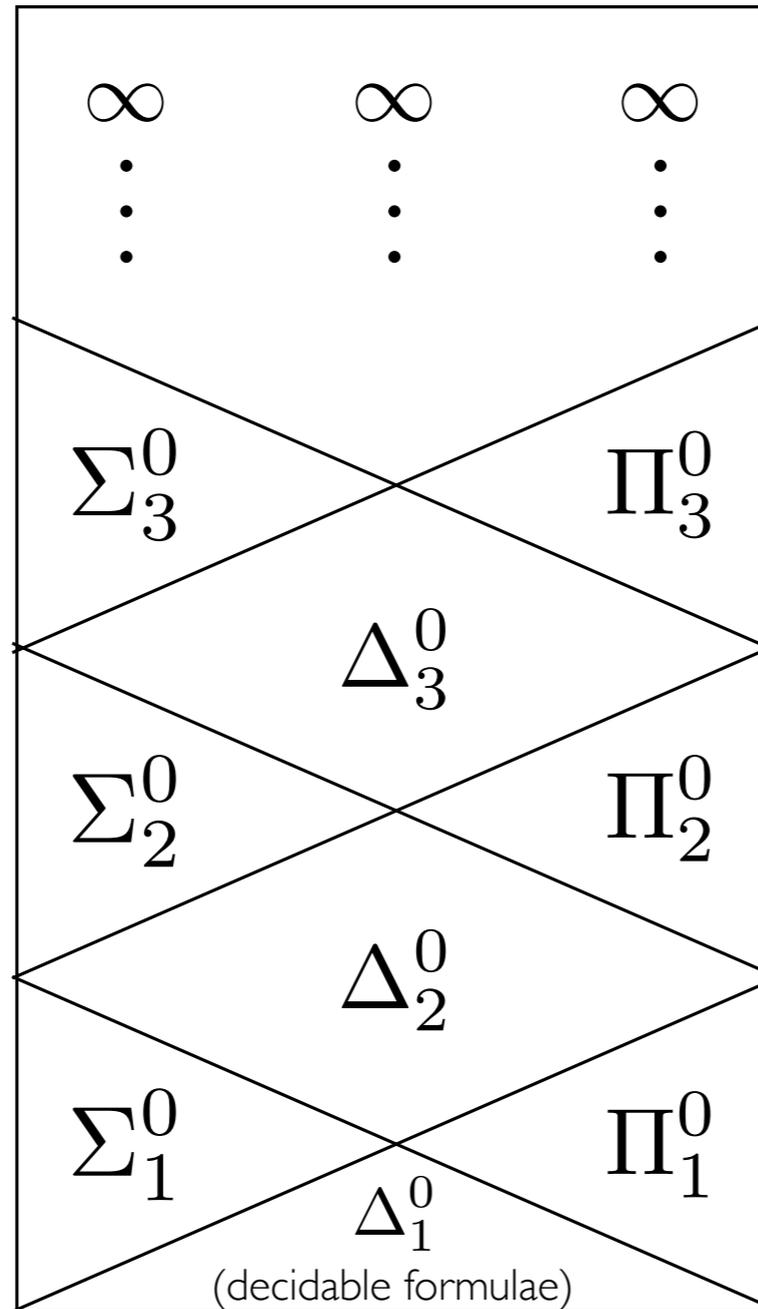
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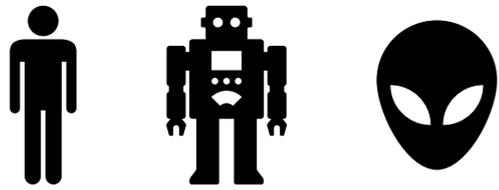
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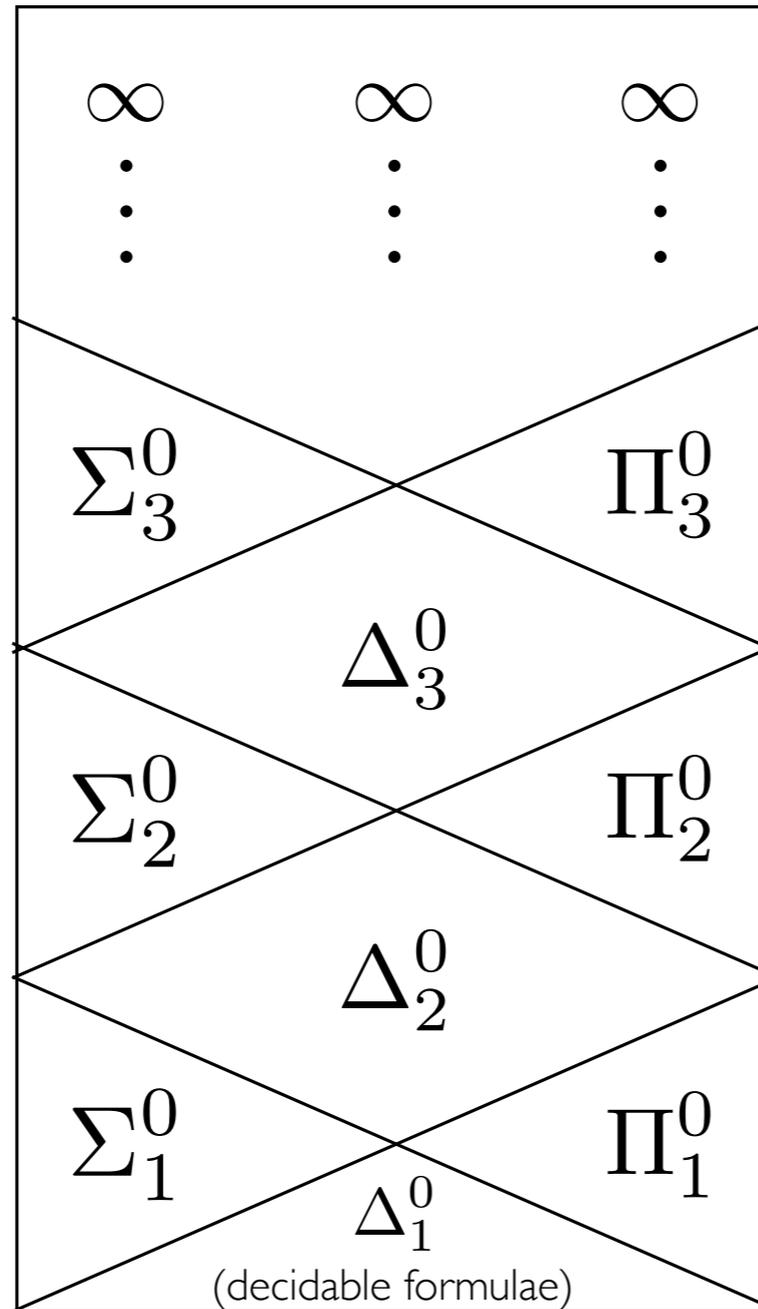
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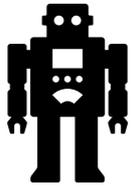
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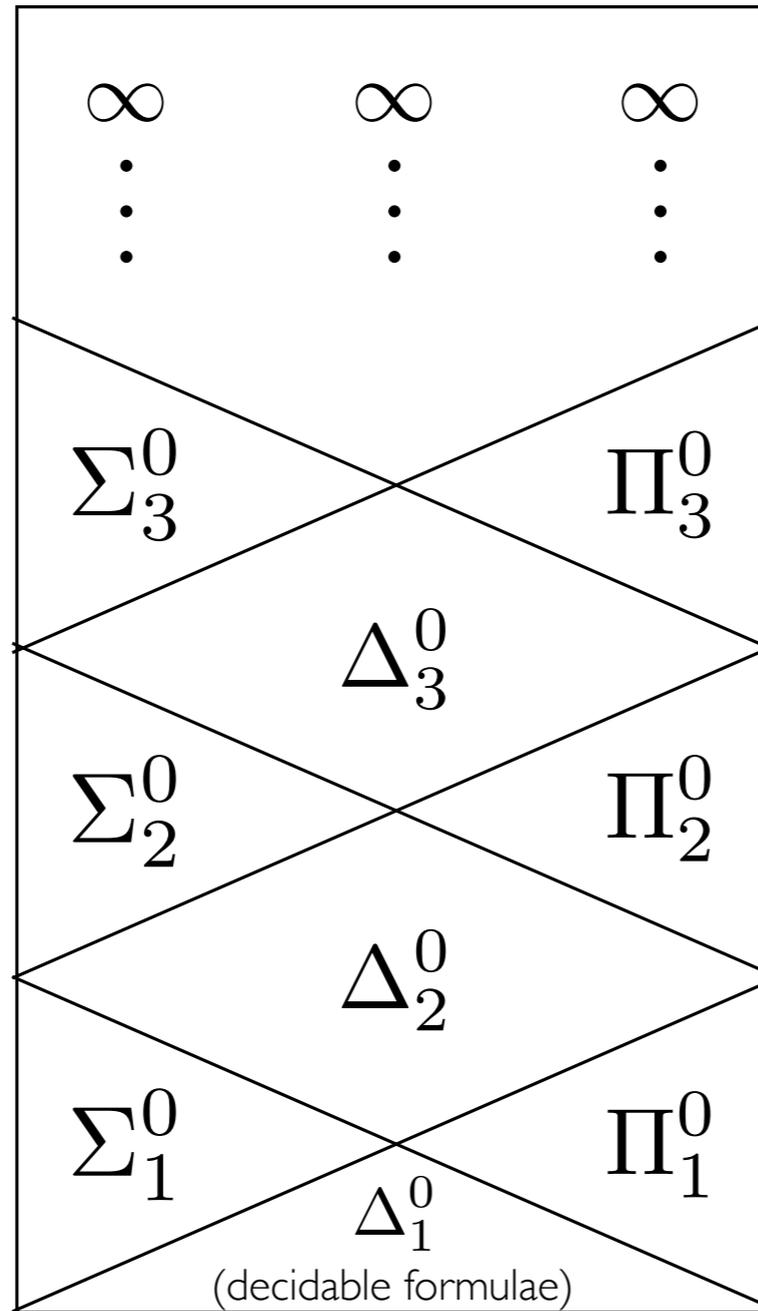
# Arithmetic Hierarchy, Part I



2SAMFUNC := { $m_1, m_2 : \forall u \forall v [\exists k (\langle m_1, u \rangle : v, k \leftrightarrow \exists k' (\langle m_2, u \rangle : v, k'))]$ }

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$\mathcal{A}^r \mathcal{H}$  (Arithmetic Hierarchy)



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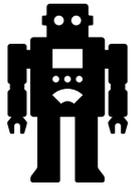
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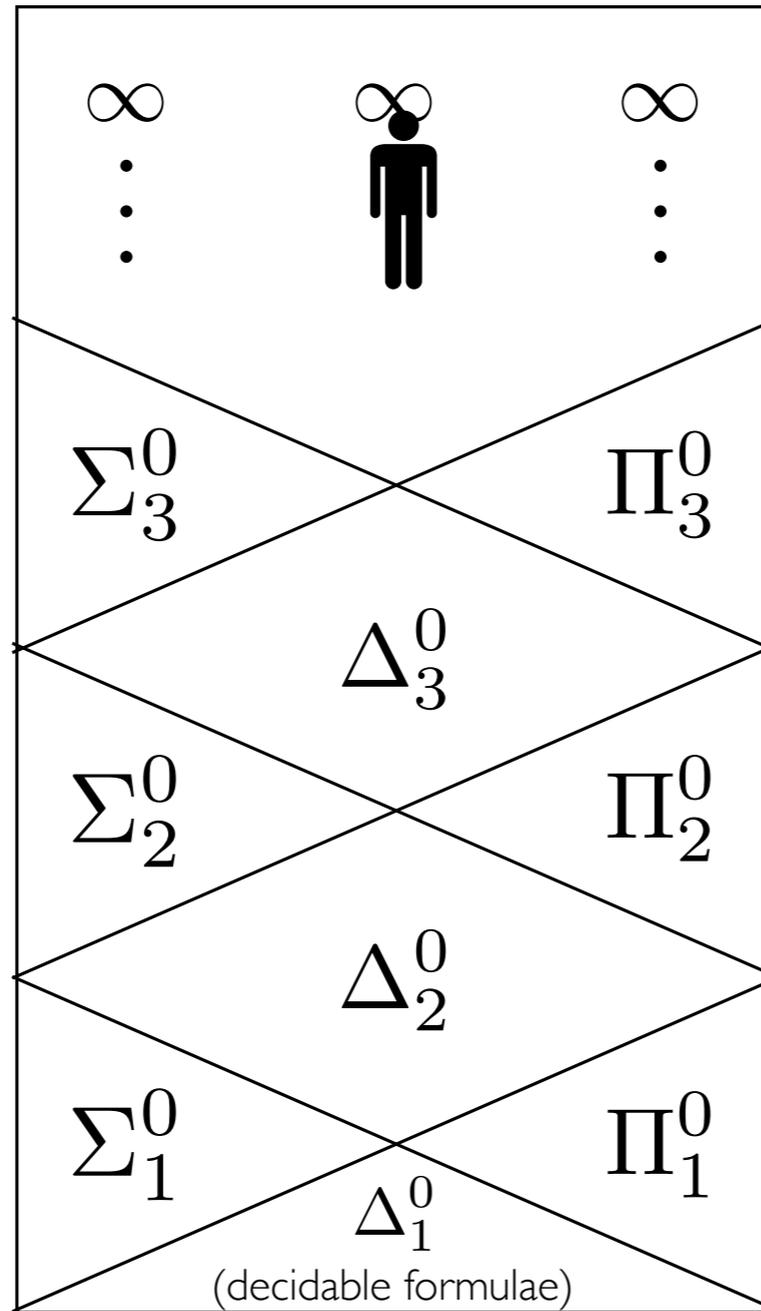
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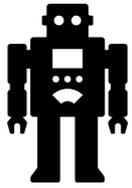
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semi-decidable



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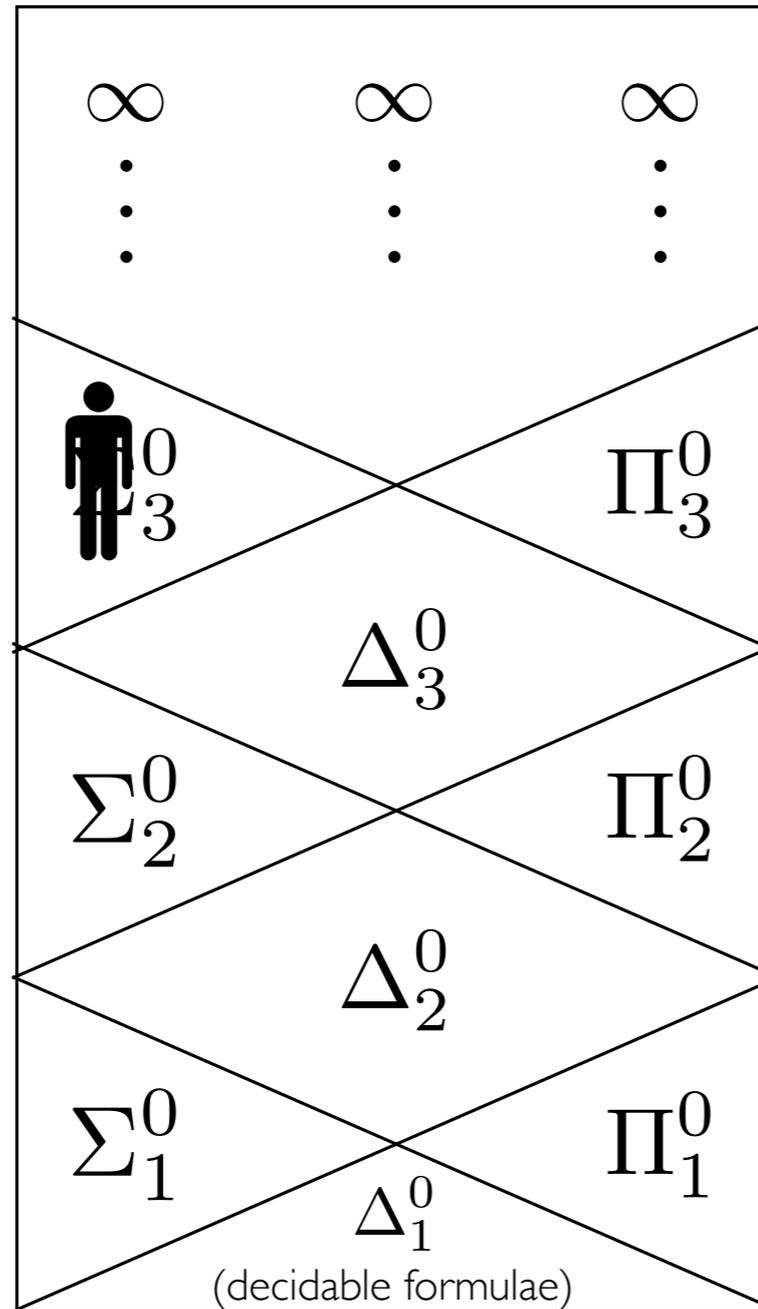
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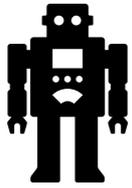
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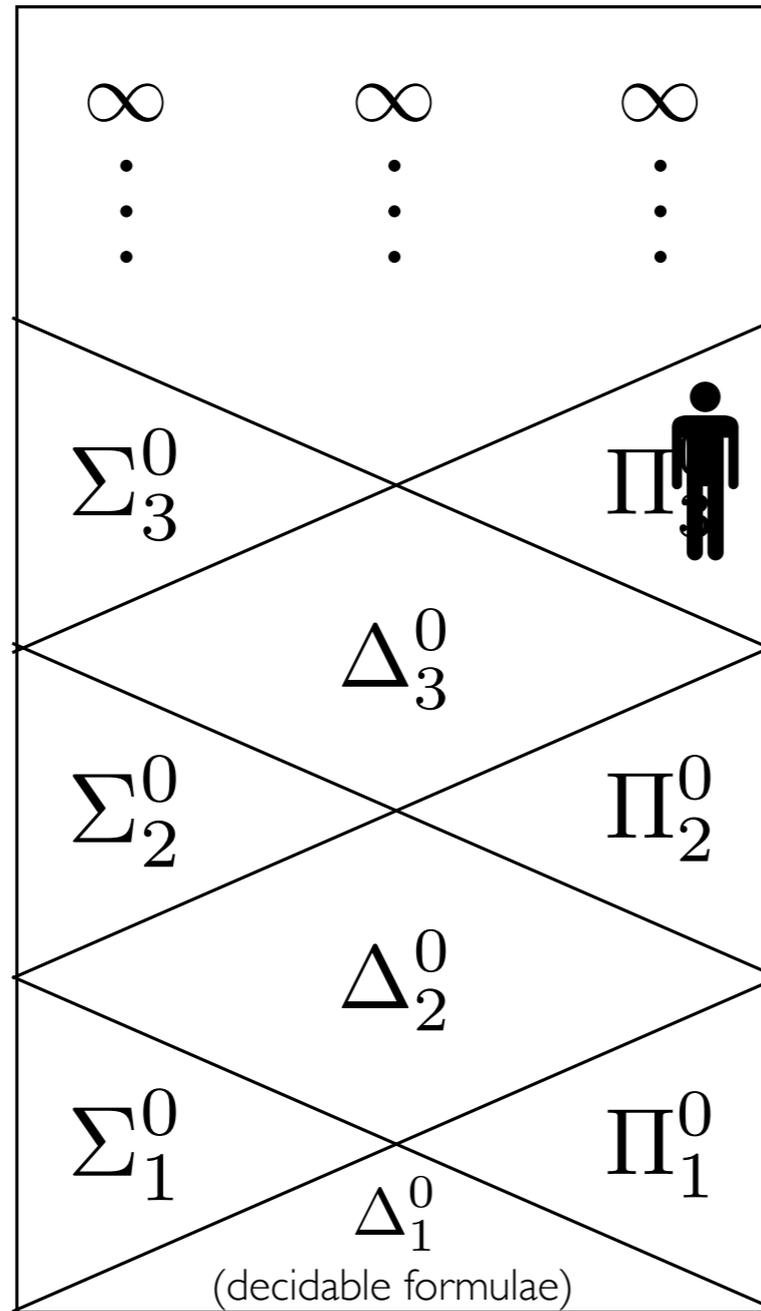
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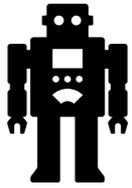
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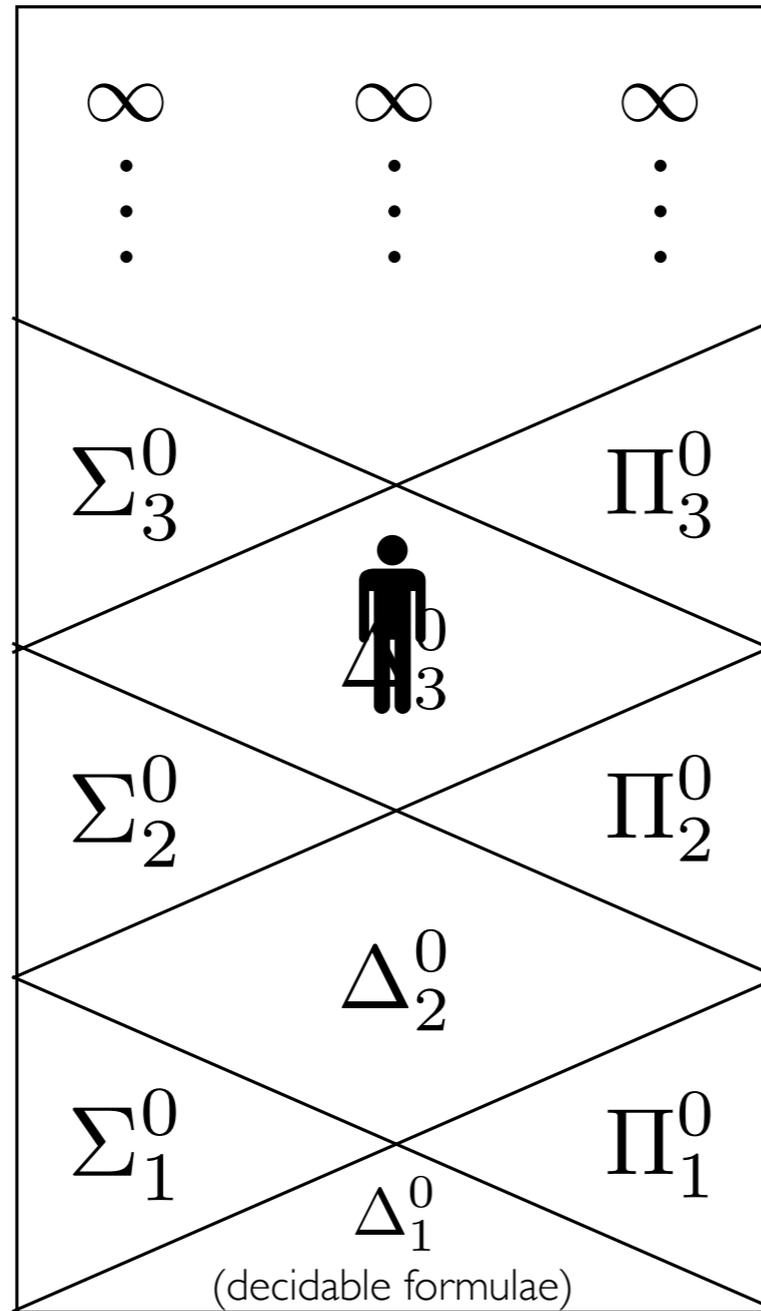
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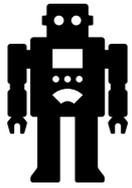
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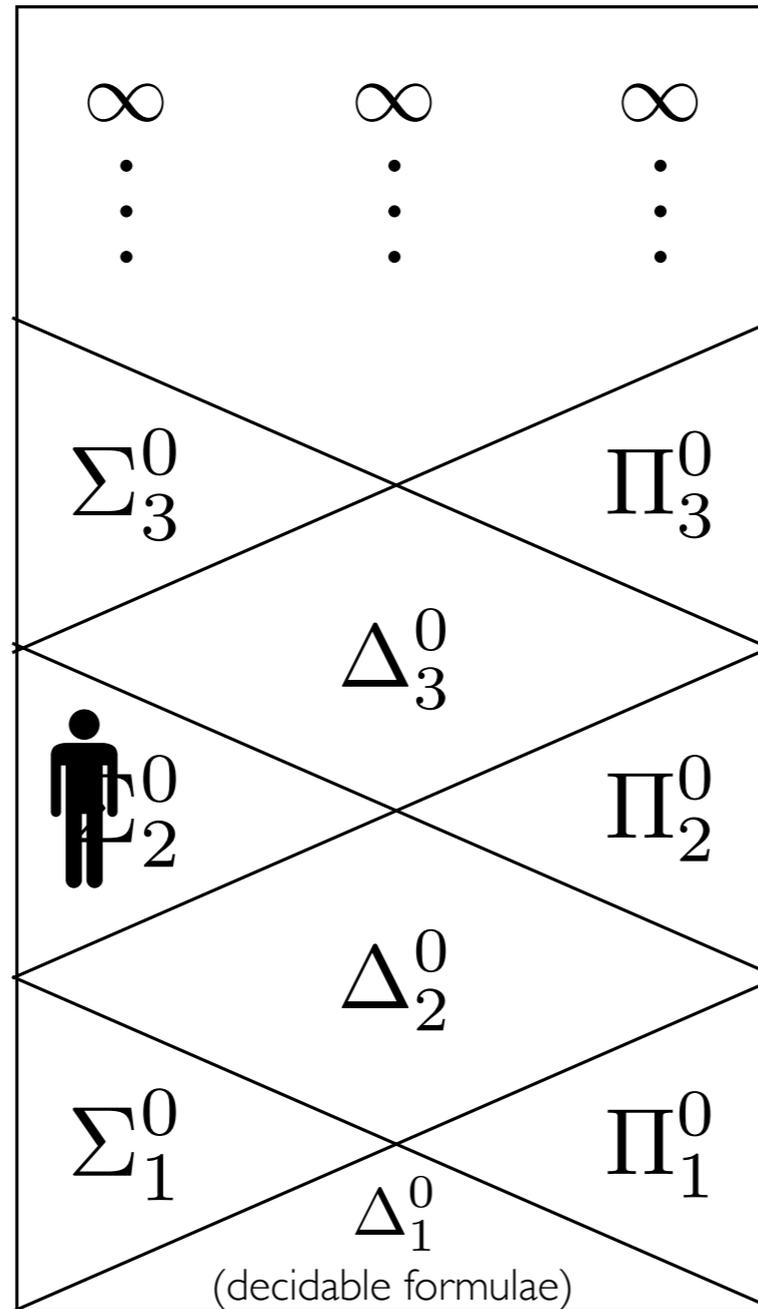
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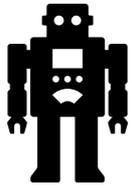
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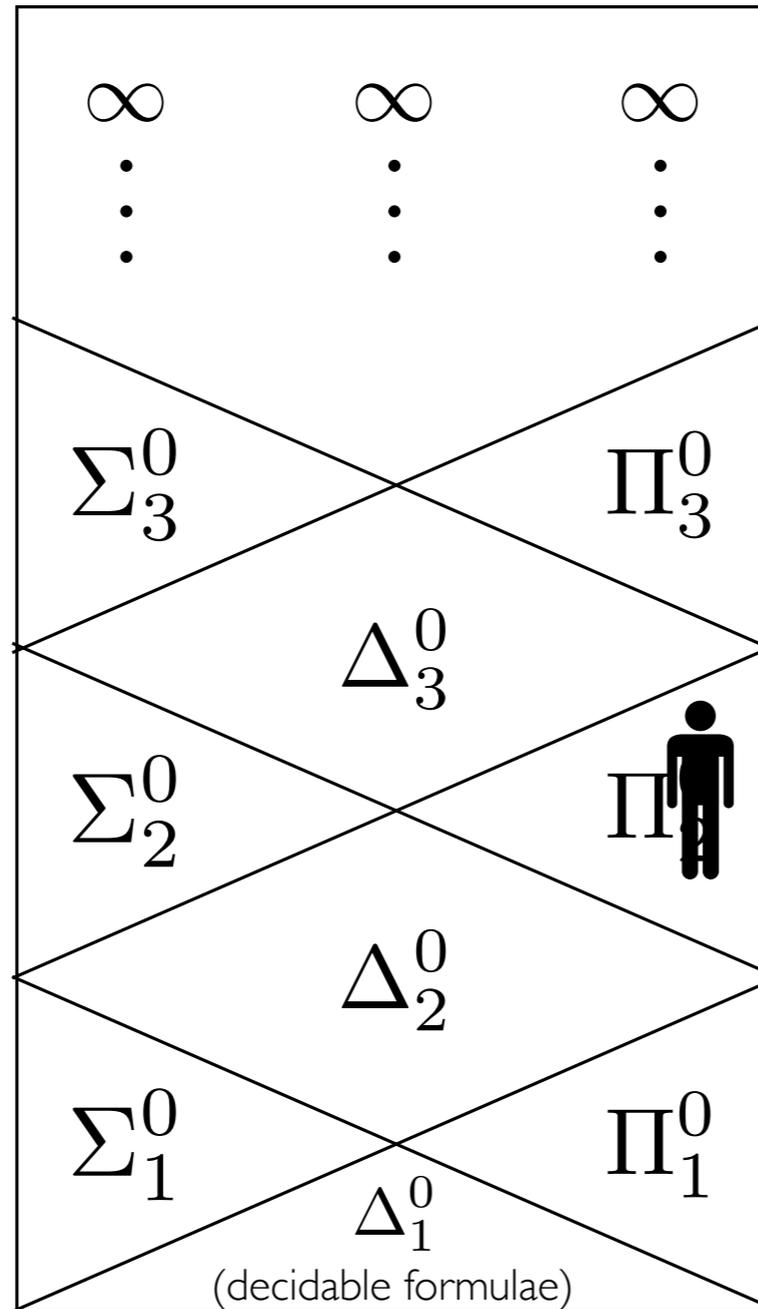
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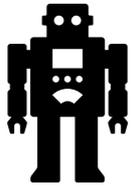
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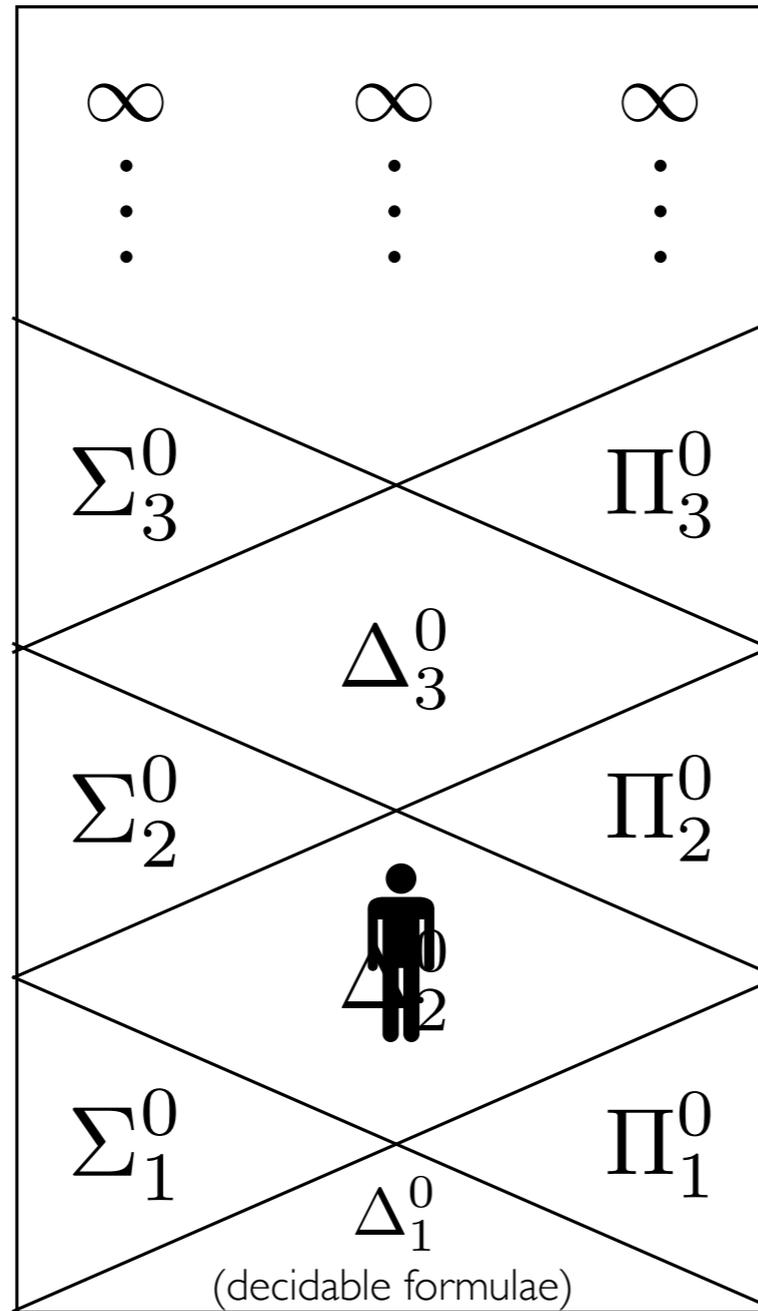
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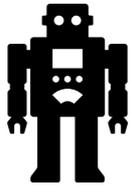
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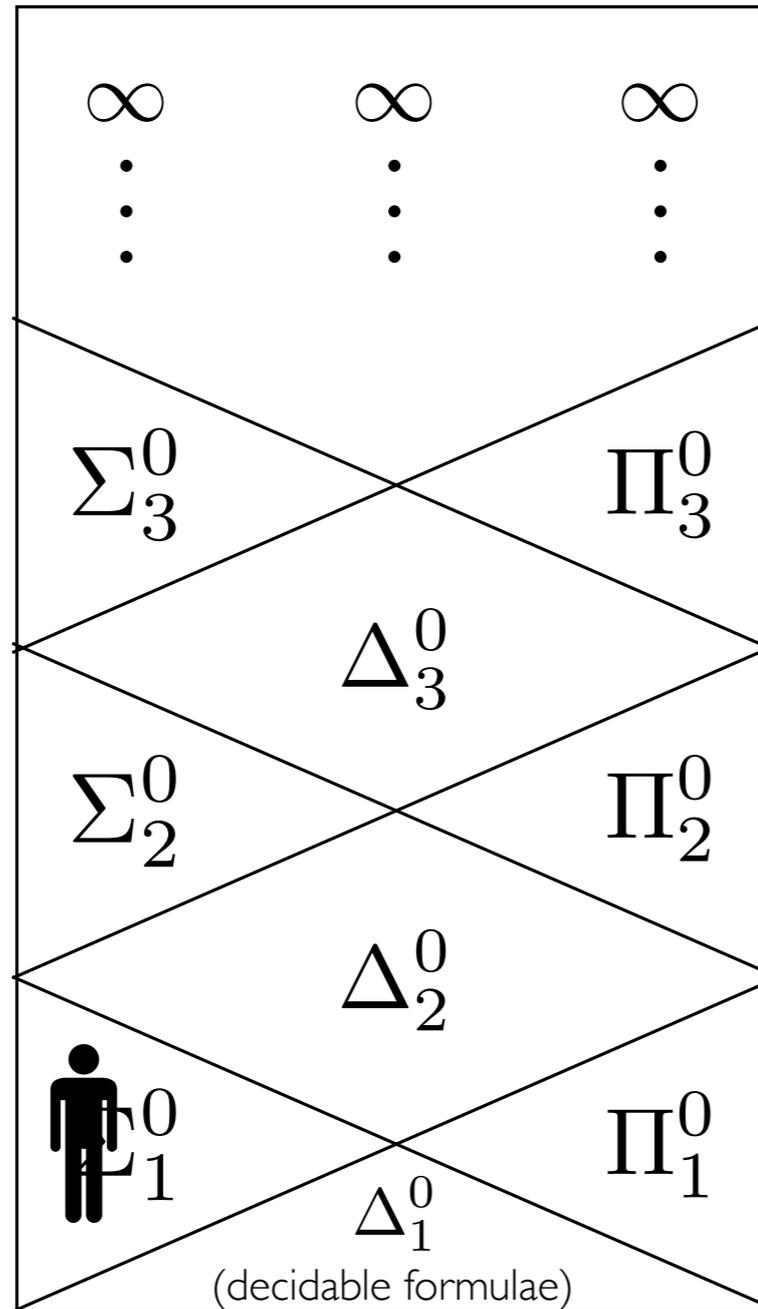
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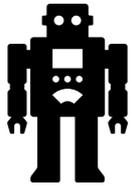
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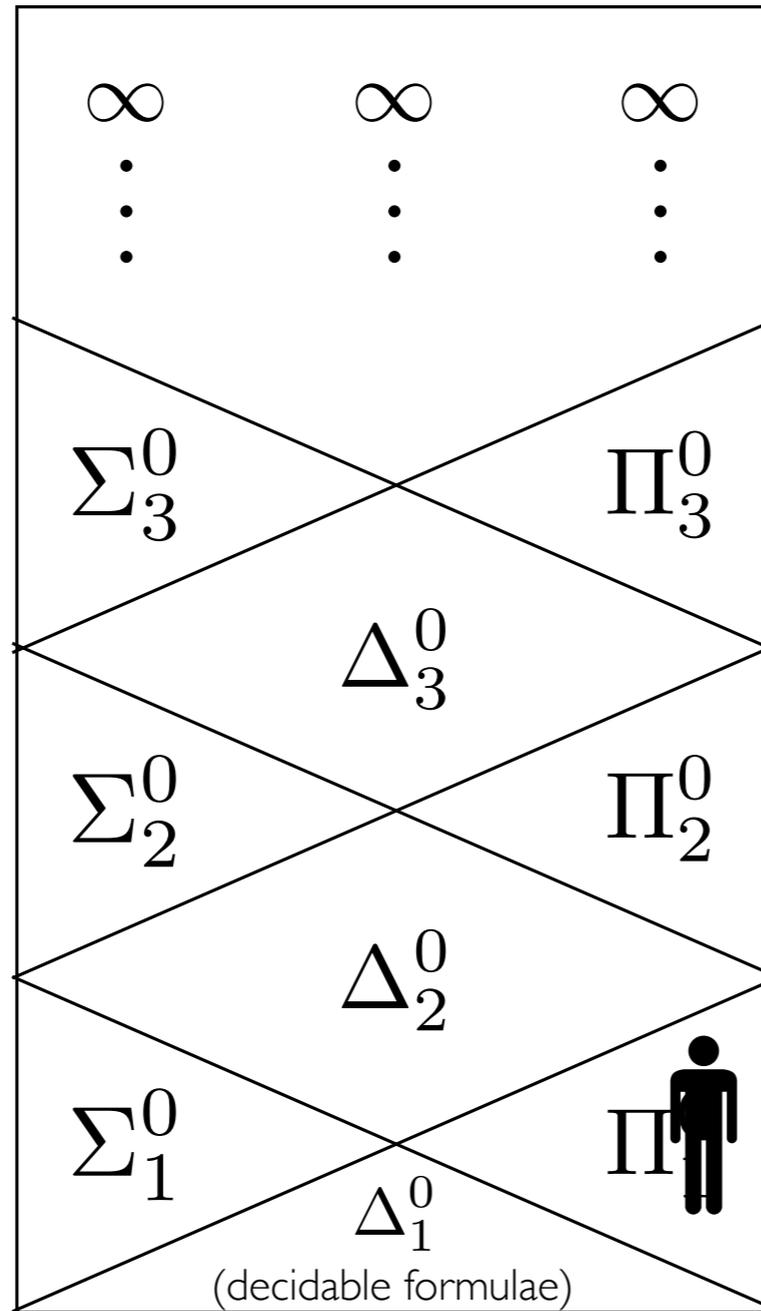
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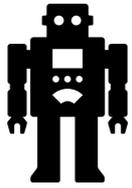
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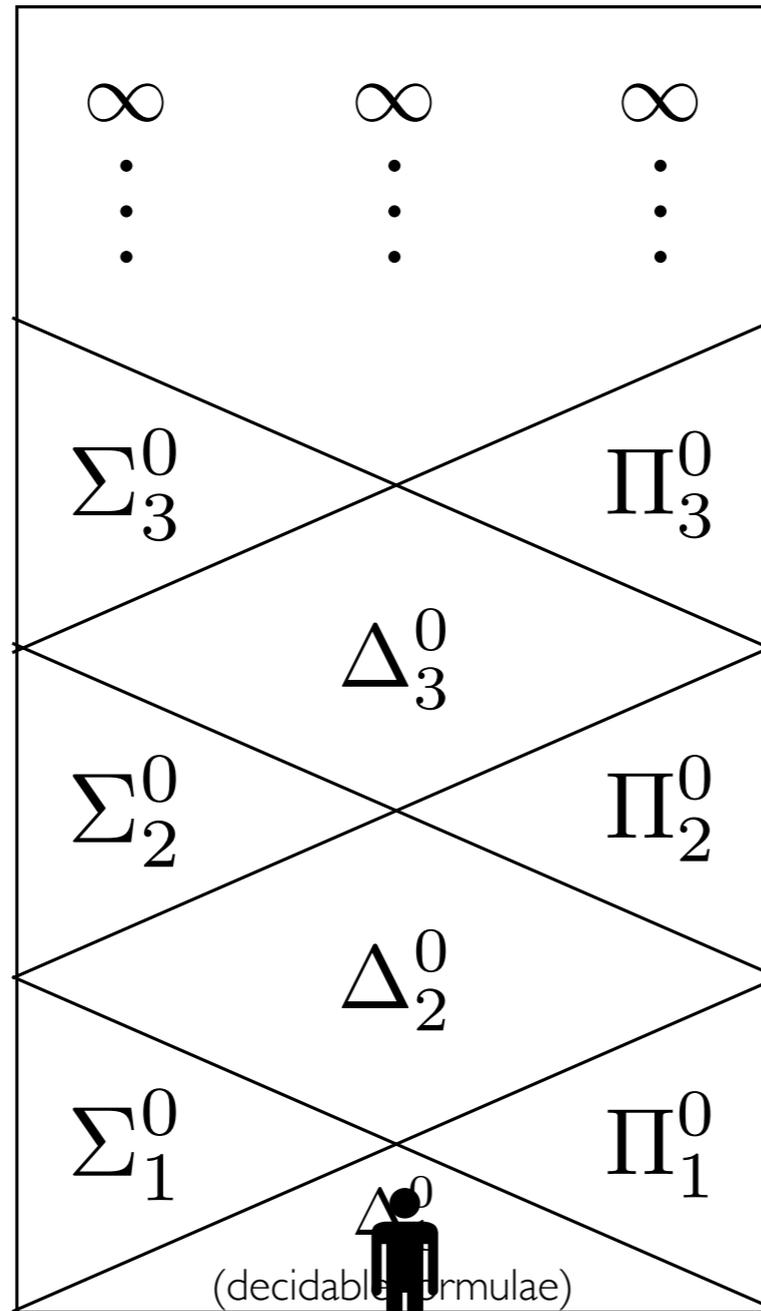
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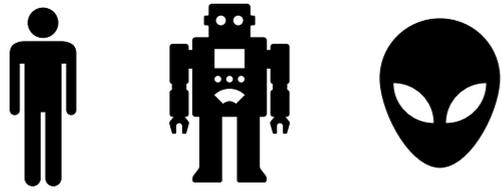
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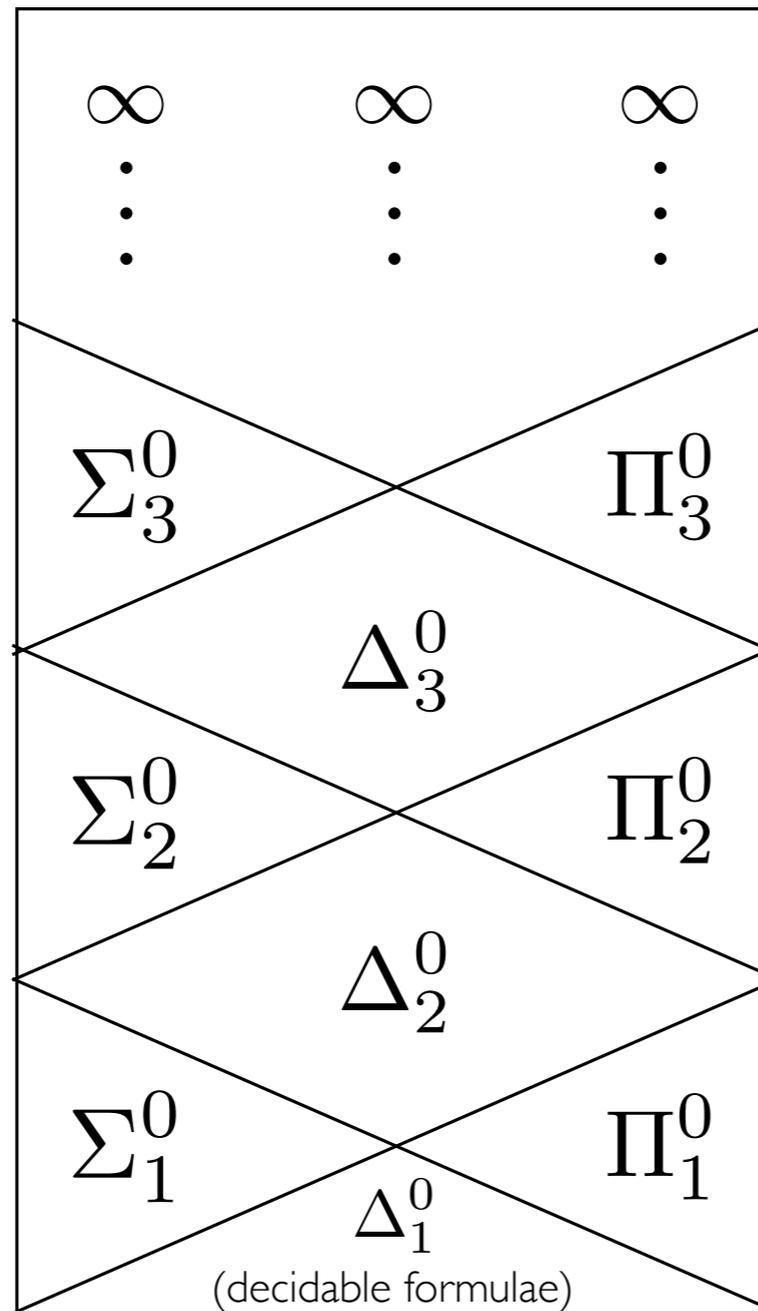
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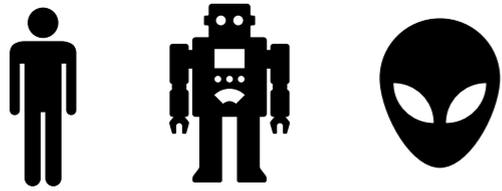
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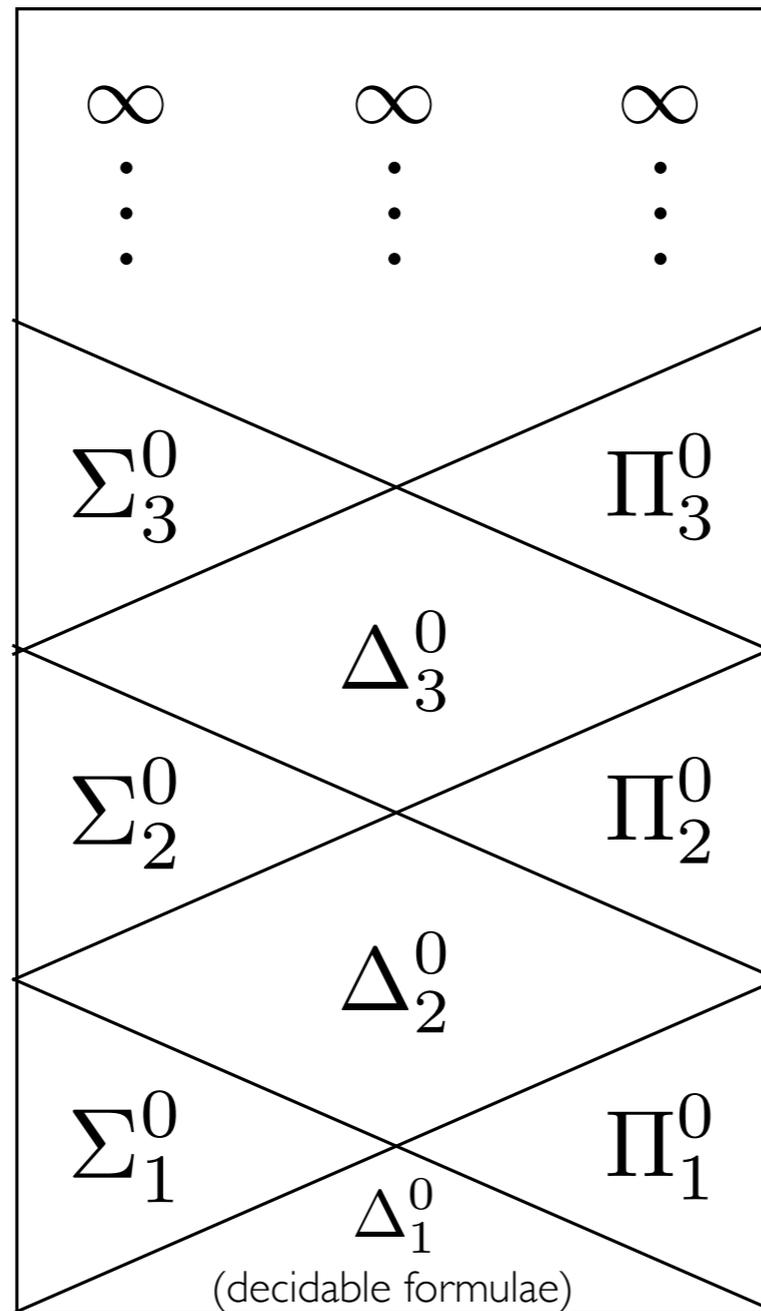
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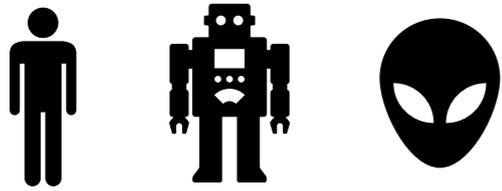
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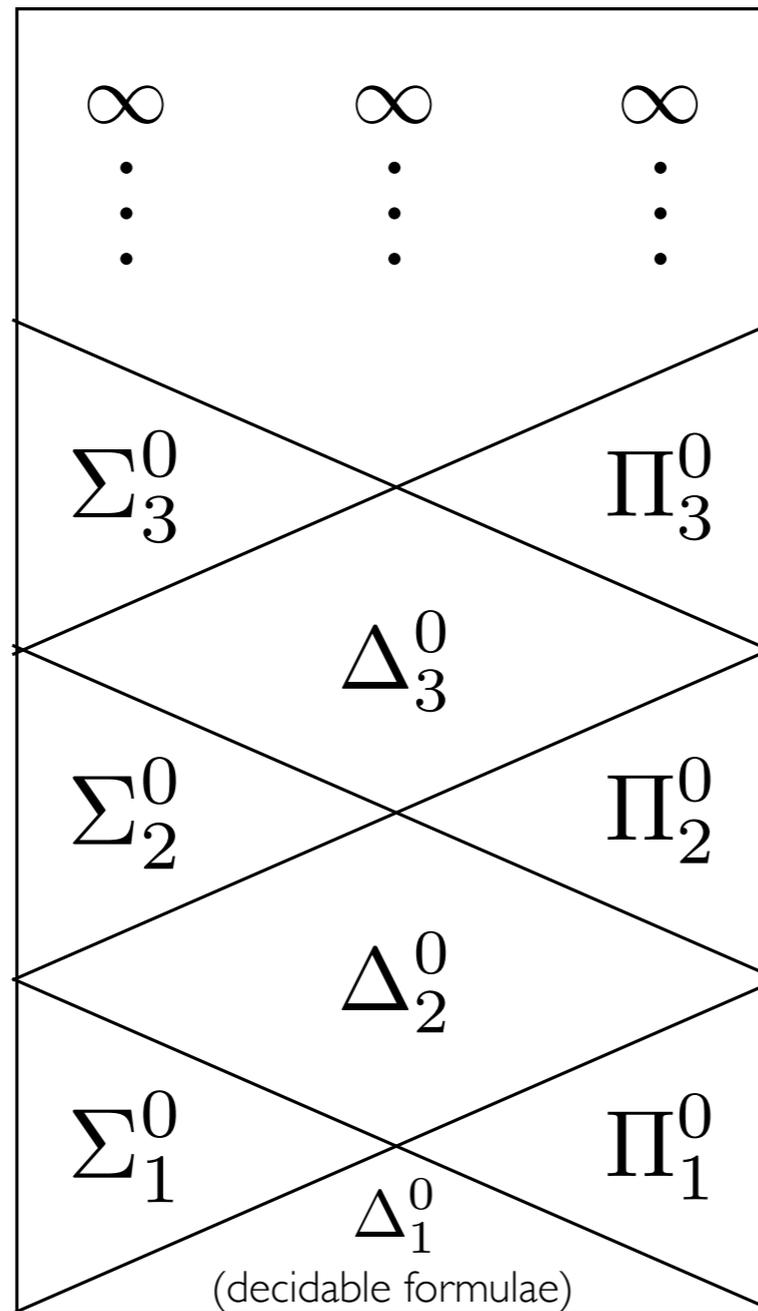
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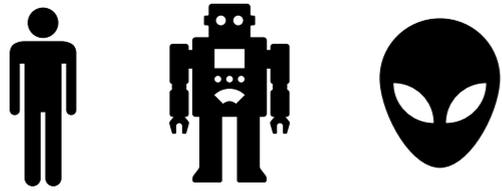
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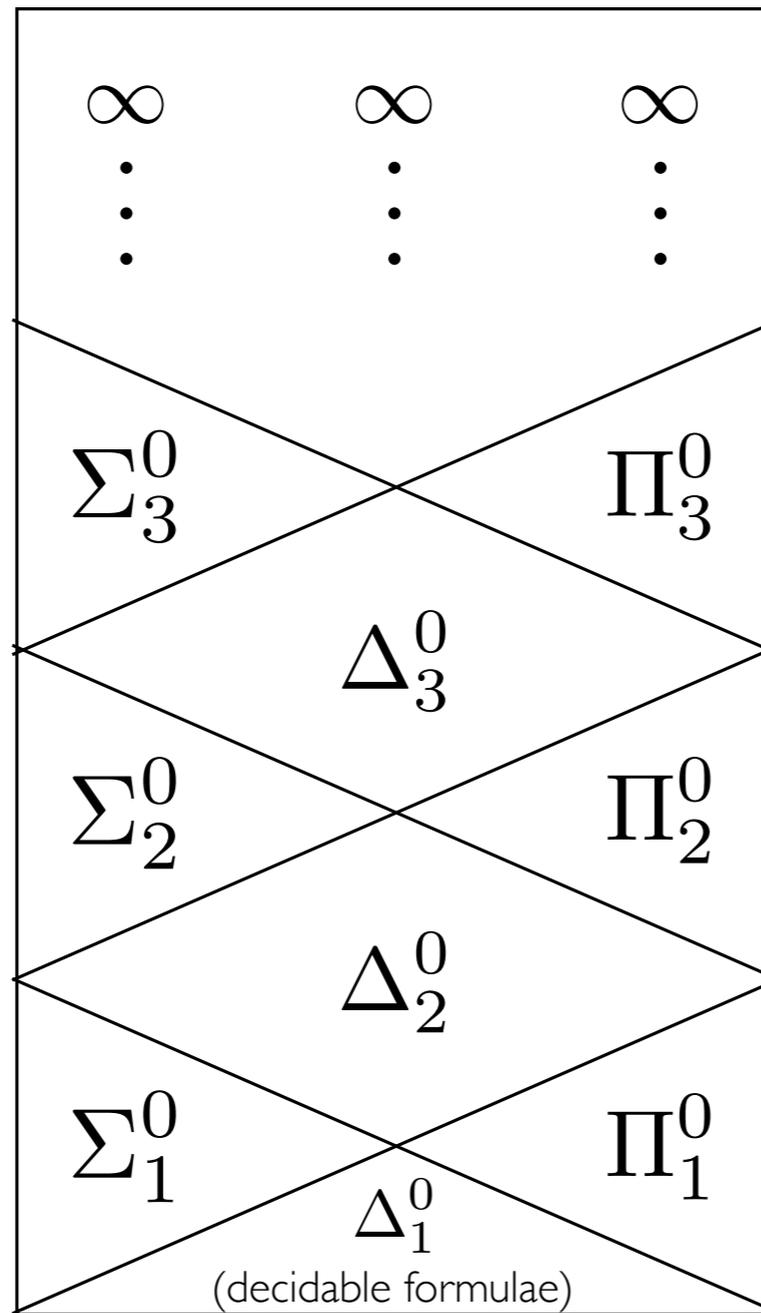
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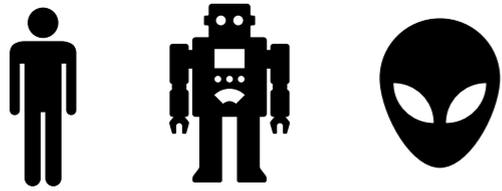
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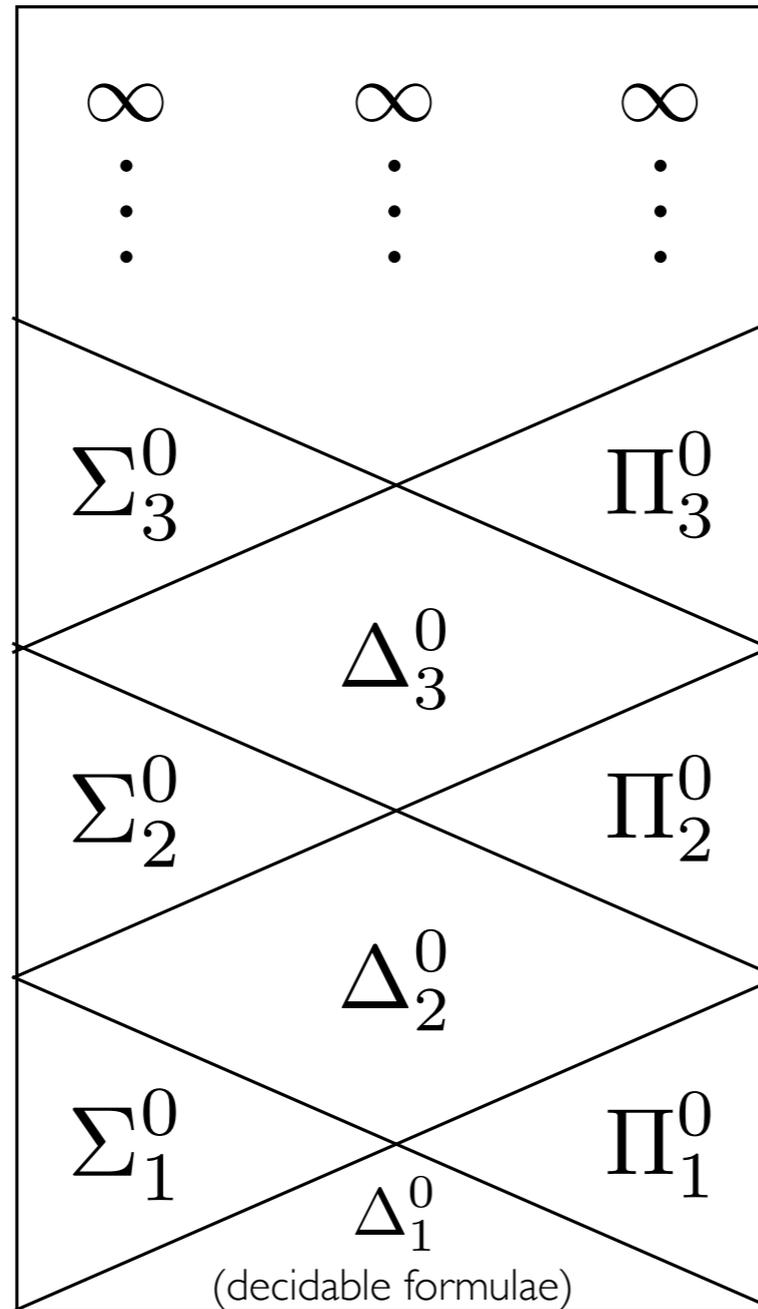
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(forall (u v) (exists (k1 k2)  
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 (comp m2 u v k2))))

semi-decidable

# Arithmetic Hierarchy, Part I

