

ONTOGENIA

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ONTOLOGY GENERATION WITH METACOGNITIVE PROMPTING IN LARGE LANGUAGE MODELS

METHODOLOGY

MOTIVATION

- **Ontology engineering is complex** and labor-intensive
- There is a need for intelligent automation **LLMs** can help with
- **Metacognitive prompting** improves performance over CoT, asking the LLM to consider its own reasoning process
- This can **reduce errors** and enhance logical consistency

CONTRIBUTIONS

- **A method** to test the effectiveness of Metacognitive Prompting and its application in automated ontology generation
- **A framework** to incorporate LLM-assisted ontology generation into the eXtreme Design methodology
- **A qualitative and quantitative evaluation** of this framework

RESULTS

Ontometrics	Case1 No pattern No MP	Case2 Pattern No MP	Case3 No pattern MP	Case4 Pattern MP	Reference Ontology
Axioms count	49	119	64	118	108
Logical axioms count	26	74	36	76	56
Class count	14	17	14	21	31
Object property count	8	11	8	14	5
Data property count	0	2	3	2	0
Properties count	8	13	11	16	5
Individual count	1	19	0	11	0
DL expressivity	ALCROI	AL(D)	ALC(D)	ALCI(D)	SRI

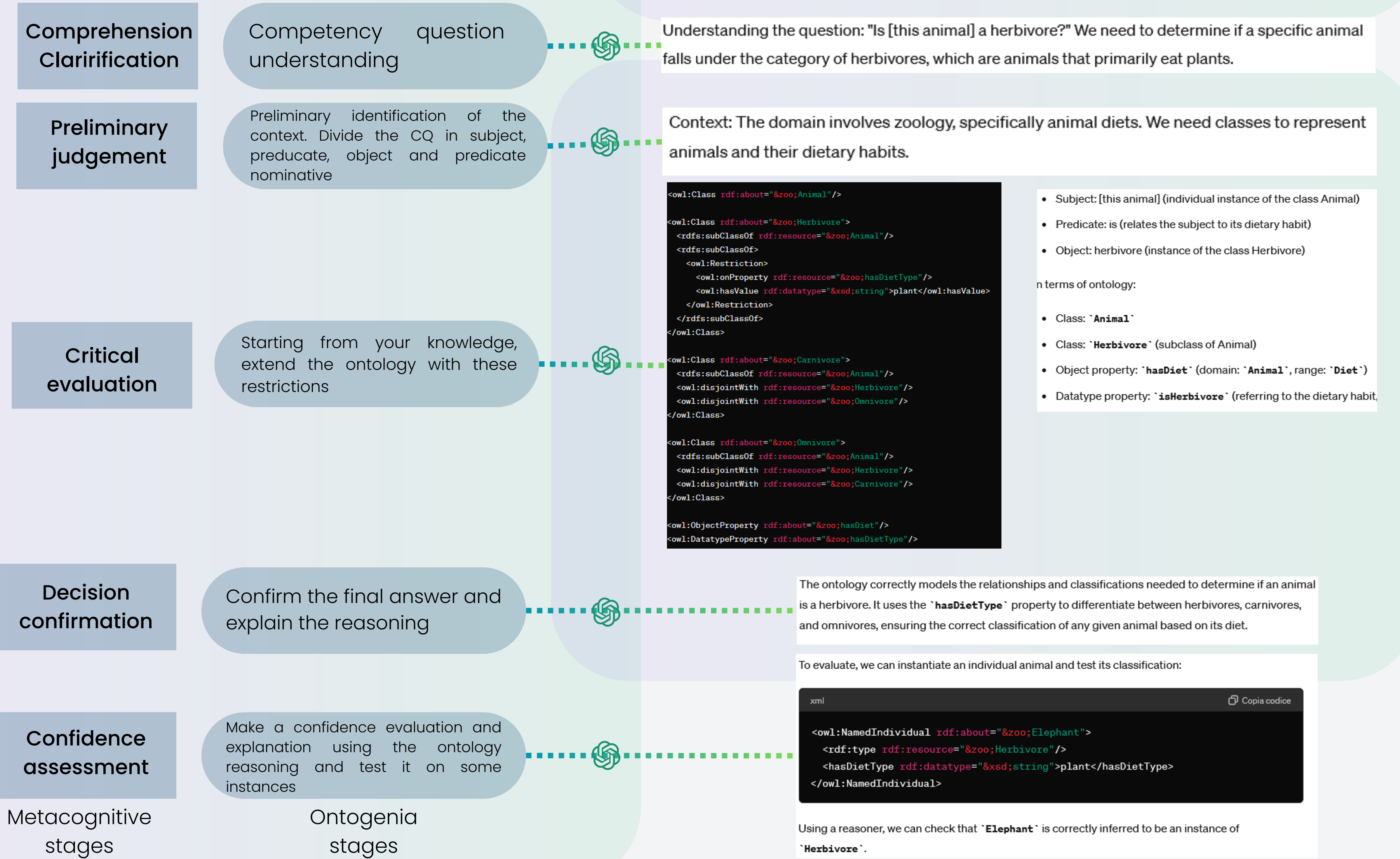
Axiom type	Case1 No pat- tern No MP	Case2 Pattern No MP	Case3 No pat- tern MP	Case4 Pattern MP
owl:Ontology	Yes	Yes	Yes	Yes
owl:Class	Yes	Yes	Yes	Yes
owl:ObjectProperty	Yes	Yes	Yes	Yes
owl:DatatypeProperty	No	No	Yes	Yes
rdfs:domain	Yes	Yes	Yes	Yes
rdfs:range	Yes	Yes	Yes	Yes
rdfs:subClassOf	Yes	Yes	Yes	Yes
rdfs:subPropertyOf	No	No	No	No
owl:disjointWith	No	No	Yes	Yes
owl:equivalentClass	Yes	No	Yes	No
owl:Restriction	Yes	No	Yes	Yes
owl:imports	No	Yes	No	Yes

- Classes and properties are successfully identified
- **Richer formalization** with MP, along with a set of different types of axiom types
- **A larger number of axioms** when a pattern is used
- A **higher number of classes and object properties**
- Low cost and time

1. Domain and CQ definition

- Use case of the African Wildlife Ontology
→ broad domain
→ covered by Ontology Design Patterns
- Comprehensive: 14 CQs

3. Procedure and prompt design



4. Evaluation measures definition

- **qualitative** ontology engineer evaluation
- **quantitative**: Ontometrics and OOPS!Ontology Pitfall Scanner

5. Testing

- **GPT-4 Turbo API** as backbone model
- Temperature 0, other parameters as default
- **Four iterations** (generic with CQs, generic with CQs and ODPs, CQs and MP, CQs with MP and ODPs) with a division of CQs in batches

```
prompt = (
    f"Read the following instructions: '{procedure}'. Basing on the procedure, and following the previous output: '{previous_output}', design an ontology that comprehensively answers the following competency questions: '{CQs}', using the following ontology design patterns: {combined_patterns}. Do not repeat classes, object properties, data properties, restrictions, etc. if they have been addressed in the previous output. Remember that crucial elements for a good ontology design are '{ontology_elements}'. When you're done send me only the whole ontology you've designed in OWL format."
)
```



LIMITATIONS

1. Absence of property hierarchical relationships
2. Lack of annotations and inverse relationships
3. With patterns, untyped class and different naming conventions are found
4. Data properties less common with patterns
5. LLMS: generated ontologies exhibit numerous intrinsic domain-related issues; pairing of classes and properties problems
6. LLMS' context limit