

# Extraction of Metaphoric Analogies from Literary Texts:

## Task Formulation, Dataset Construction, and Evaluation



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### 4-term metaphoric analogies

#### Metaphoric mapping

- **Target domain**: semantic domain of the concepts **used literally** in the analogy.
- **Source domain**: semantic domain of the concepts **used metaphorically**.

#### Structured metaphoric analogy

**T1** is to **T2** what **S1** is to **S2**

- **Explicit terms** occur in the text.
- **Implicit terms** do not occur in the text, but are conveyed by the reader for understanding the analogy.

i : number of implicit terms  
<> : implicit terms

Metaphoric analogies in literary texts	T1	T2	S1	S2	i
<b>Books</b> are like <b>imprisoned souls</b> till someone takes them down from a <b>shelf</b> and frees them. (S. Butler)	books	shelf	imprisoned souls	<prison>	1
<b>Money</b> is the <b>mother's milk</b> of <b>politics</b> . (J. Unruh)	money	politics	mother's milk	<baby>	1
An election is coming. Universal peace is declared, and the <b>foxes</b> have a sincere interest in prolonging the lives of the <b>poultry</b> . (G. Eliot)	<candidates>	<voters>	foxes	poultry	2

- **Task validation**: 5 annotators with background in linguistics or metaphor studies extract analogical frames from 20 short texts given the term T1.
- Average pairwise inter-annotator agreement scores with a **Cohen's Kappa score of 0.68 for the frames containing explicit terms**
  - **metric**: lemmatized head-noun match

**Dataset construction**: 204 short texts are selected from existing collections of literary metaphors and labelled

### Task

#### Input :

- a short text
- a term T1, T2, S1 or S2

#### Instruction :

- Extract the other explicit terms forming the 4-term metaphor
- Generate eventual missing implicit terms

#### Output :

The structured metaphoric analogy : values of the 4 frames **T1**, **T2**, **S1** and **S2**

### Experiments

#### 7-shot in-context learning with 5 models

- Llama-3, Mixtral 8\*7 & Mixtral 8\*22
- GPT-3.5 & GPT-4

#### Dataset instances are tested with

- Each of the 4 frames given as input
- 3 different example sets in the prompt

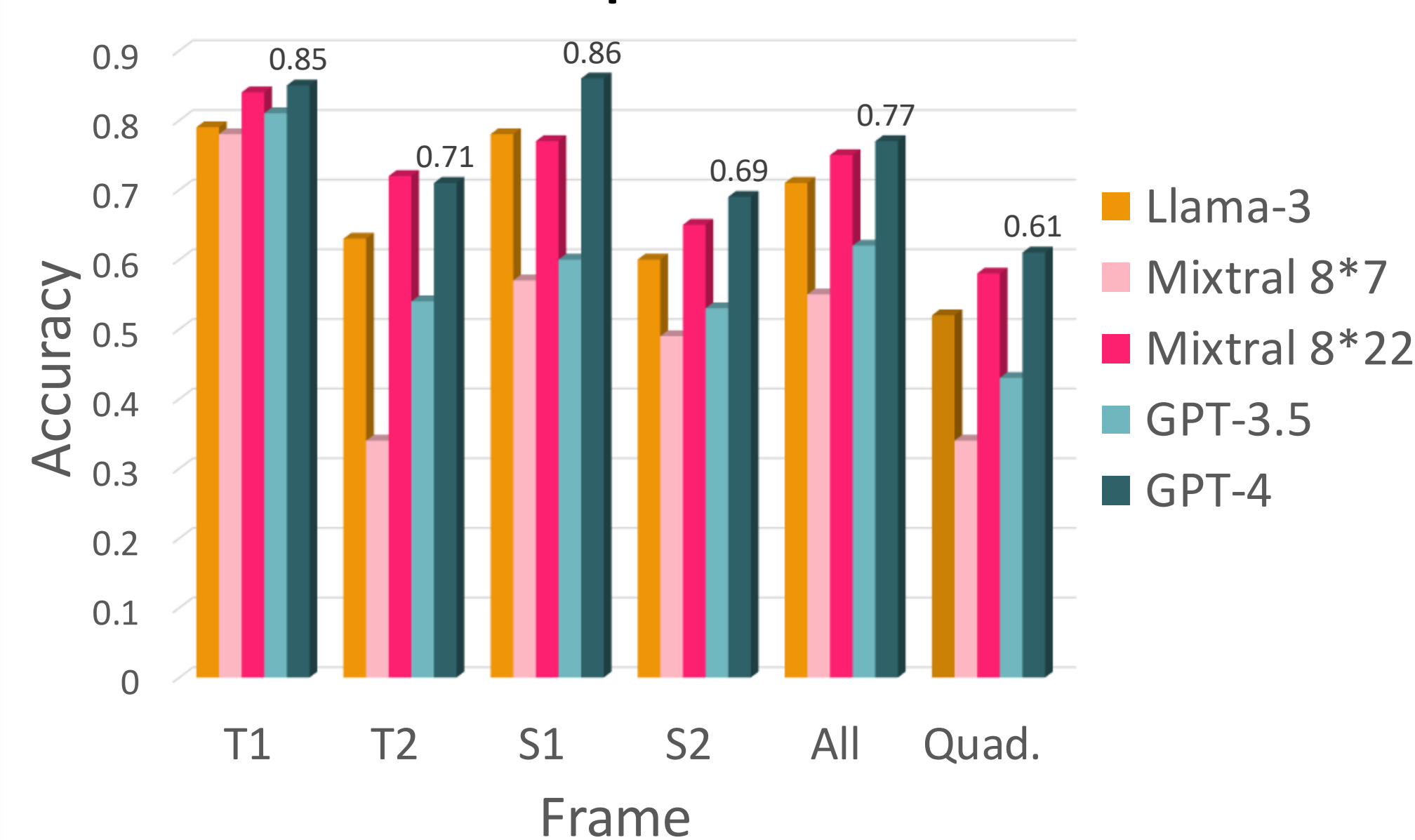
### Evaluation of the extracted explicit terms

#### Metric: Lemmatized head noun

**All** : correct when any frame T1 T2, S1 or S2 is correctly extracted (669 distinct explicit terms)

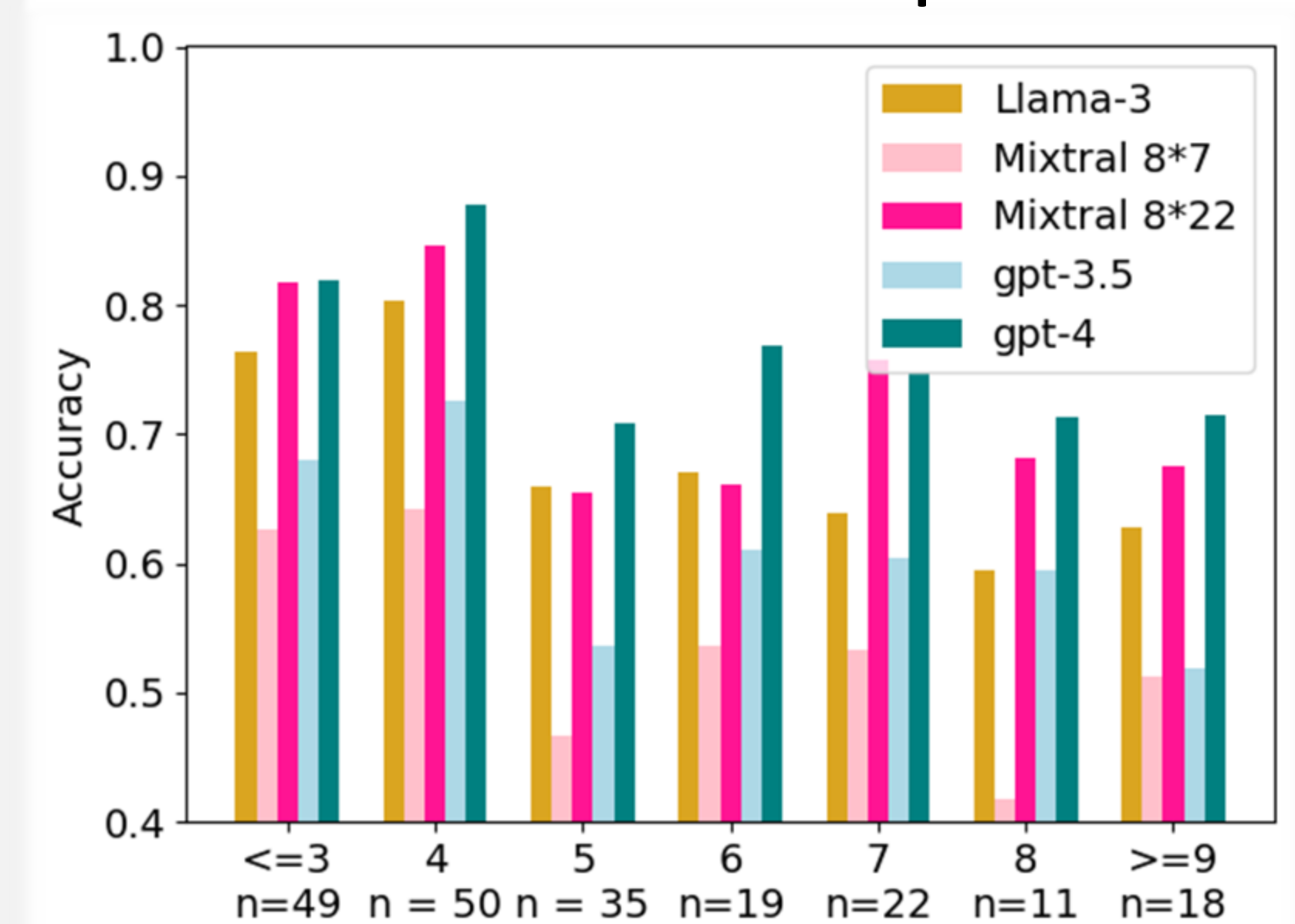
**Quad.** : correct when all the explicit terms of a quadruple are correctly extracted (204 distinct quadruples)

Accuracy results for explicit terms split by output frame



n : number of short text in each bucket

Accuracy results for explicit terms split by number of nouns in the input text



### Evaluation of the generated implicit terms

**Manual evaluation**: 2 annotators rate the generated implicit terms of 50 different examples, where explicit terms were extracted correctly.

**Example** : **Written laws** are like **spiders' webs**, and will, like them, only entangle and hold **the poor and weak**.

**Explicit terms**: **T1** : **written laws**  
**S1** : **spiders' webs**  
**T2** : **the poor and weak**

**Suggested implicit term**  
**S2** : **<weak insect>**

Model	Generated S2 terms	Rating
Mixtral 8*22	spiders	0
	entangle and hold	0
GPT-4	entangled beings	1
	small insects	2

0 : incorrect  
1 : imperfect  
2 : very good

- **Sentence score**: average rating of all generated terms
- **Model score**: average sentence score

#### Score of the best two models:

Model	Total	Deduplicated	Score
Mixtral 8*22	356	153	0.75
GPT-4	369	188	1.21

- **Total** : total number of evaluated terms
- **Deduplicated** : number of distinct generated words

### Conclusion

- Language models can become valuable tools for the conversion of unstructured metaphors to structured analogical concept mapping.
- The performance of the best models analyzed, namely GPT-4 and Mixtral 22\*8 are in line with human annotators.

#### Possible extensions of the analogy framework

- n to n concepts mapping extraction
- Relation extraction

#### Questions for future work

- How robust are the models for analogy extraction on open longer texts?
- What are the type of metaphoric analogies that the models struggle the most to extract?

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