

Ingegneria dei dati
Homework 4
(da svolgere in gruppo)

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Homework 4: knowledge extraction

- From the collection of scientific papers that your team has downloaded (then, all the papers refer to the same topic), randomly select about 10 papers, each containing around 3 tables each
- The final dataset for this experiments should be composed by more than 30 tables

Homework 4: knowledge extraction

- Extract the claims presented in the tables and in their associated context (references, caption, footnotes).
- Claims must be extracted according to the following format:
 - |{**Specification, Specification, ...**}, **Measure**, **Outcome** |
 - Specification: |**name, value**| pair describing the details of an experiment
E.g.: |dataset, Spider|
 |LLM, Llama27b|
 - Measure: metric or measure used to evaluate the experiment
E.g.: F1-measure
 - Outcome: outcome value related to metric
E.g.: 0.89

Example – Claims extraction (Result table)

Paper: [Enhancing Text-to-SQL Translation for Financial System Design](#)
(paper cs topic: text2sql) (paper id: 2312.14725)

Model Type	Model Name	Parameter Size	Level 1	Level 2	Level 3	Level 4	All
General LLM	ChatGPT-3.5-turbo	175B	0.760	0.799	0.408	0.493	0.623
	DIN-SQL+GPT-4	1.76T	0.861	0.866	0.700	0.654	0.762
	CodeX-Davinci-3	175B	0.730	0.799	0.392	0.382	0.570
	MPT-7B-instruct	7B	0.262	0.381	0.117	0.091	0.205
	ALPACA	7B	0.311	0.460	0.192	0.083	0.242
	KOALA	7B	0.195	0.218	0.017	0.071	0.131
	OpenAssistant-pythia	12B	0.202	0.322	0.025	0.069	0.157
	ORCA-mini	7B	0.243	0.280	0.101	0.076	0.169
Code Specific LLM	LLaMA-2	7B	0.225	0.393	0.101	0.081	0.192
	CodeGen2	7B	0.375	0.498	0.167	0.066	0.257
	Starcoder	15.5B	0.584	0.628	0.275	0.208	0.410
	Vicuna	7B	0.060	0.134	0.008	0.042	0.064
	nsql	6B	0.772	0.732	0.608	0.277	0.548
Seq-to-Seq Model	T5(tscholak/cxmefzzi)	3B	0.828	0.782	0.650	0.434	0.641
	PICARD+T5	3B	0.790	0.799	0.558	0.502	0.652
	RESDSQL	3B	0.872	0.857	0.666	0.696	0.775

Paragraph (reference)

In our experimentation, we organized the models into three distinct groups as illustrated in Table 1: general purpose LLMs, Code-Specific LLMs, and Sequence-to-Sequence models. Table 1 further presents the Execution Match score on the SPIDER dataset for each studied LLM and for each of the four difficulty levels. Note

Table 1: Benchmark Results of Execution Match of all Models we tested on the "dev" SPIDER dataset

1. |{|Model Type, General LLM|, |Model Name, ChatGPT-3.5-turbo|, |Parameter Size, 175B|, |Dataset, Spider dev|, |Difficulty Level, 1|}, Execution Match , 0.760|

(claims)

2. ..

Example - Discussion

1. |{|Model Type, General LLM|, |Model Name, ChatGPT-3.5-turbo|, |Parameter Size, 175B|, |Dataset, Spider dev|, |Difficulty Level, 1|}, Execution Match , 0.760|
2. ..

- Specifications

- |*Model type, GeneralLLM*| is located in header and index
- |*Parameter size, 175B*| is located in header and cell
- |*Dataset, Spider dev*| is located in caption
- |*Difficulty Level, 1*| must be inferred from text and table header

- Measure: metric or measure used to evaluate the experiment

- *Execution Match* metric was located caption and text, but not mentioned in the table

- Outcome: outcome value related to metric

- Most cells of the table report outcomes but not all of them; plus, their metric is not directly mentioned in the table, but rather in the caption and the text

Example - Claims extraction (Data table)

- [Paper](#)

Table and caption:

TABLE IV: Partition characteristics

Dataset	# cutting edges	α
LUBM-8000	23,624,351	1.23
LUBM-20480	61,518,672	1.21
SNIB-15000	58,823,356	1.52

Paragraph:

We adopted $n = 500$, i.e. the RL-graph of each dataset was partitioned into 500 subgraphs using METIS. The effects of \mathbb{P}_{METIS} and \mathbb{P}_{I-UHC} is given in *Table IV*, manifested as the number of cutting edges and *replication factor* α respectively.

Dataset, LUBM-8000 |, | RL-GRAPH partitions, 500 subgraphs |, | # cutting edges, 23,624,351 |, | replication factor alpha, 1.23 |}

Dataset, LUBM-20480 |, | RL-GRAPH partitions, 500 subgraphs |, | # cutting edges, 61,518,672 |, | replication factor alpha, 1.21 |}

Dataset, SNIB-15000 |, | RL-GRAPH partitions, 500 subgraphs |, | # cutting edges, 58,823,356 |, | replication factor alpha, 1.52 |}

Table Classes (based on structure)

- **Four** different table **classes**:
 - Relational
 - Nested relational (*tabelle relazionali nidificate*)
 - Cross-table
 - Nested cross-table (*cross-table nidificate*)

Example Table Classes

- Relational

Model	Parameters	Precision	Recall	F1	
Llama 3.2	7B	x	y	z	
Gemma	70B	x2	y2	z2	
Mixtral	80B	x3	y3	z3	

Example Table Classes

- Nested Relational

Model Type	Model Name	Parameter Size	Level 1	Level 2	Level 3	Level 4	All
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	MPT-7B-instruct	7B	0.262	0.381	0.117	0.091	0.205
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	Starcoder	15.5B	0.584	0.628	0.275	0.208	0.410
	Vicuna	7B	0.060	0.134	0.008	0.042	0.064
	nsql	6B	0.772	0.732	0.608	0.277	0.548
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	PICARD+T5	3B	0.790	0.799	0.558	0.502	0.652
	RESDSQL	3B	0.872	0.857	0.666	0.696	0.775

Table 1: Benchmark Results of Execution Match of all Models we tested on the "dev" SPIDER dataset

Example Table Classes

- Cross-table

	D1	D2	D3		
M1	0.9	0.7	0.6		
M2		
M3	0.7		
Caption: Accuracy results of methods M1, M2, M3 on datasets D1, D2, D3					

Example Table Classes

- Nested cross-table

		Textual		Numeric	
		D1	D2	D3	
Neural	M1	0.9	0.7	0.6	
	M2	
Graph	M3	0.7	
Caption: Accuracy results of methods M1, M2, M3 on datasets D1, D2, D3					

Task 1: Claim Extraction

- Use html, caption, references and footnotes for the extraction process.
- *More specifications the better*
- Extract also a specification called |Task, ...|, which represent the target task (e.g.: |task, record linkage|)
- Produce the ground truth for each table

Task 1: File and Claims Formats

- For each pair (paper, table) you need to produce a **json file** named paperID_tableID_claims.json containing the set of the extracted claims.
 - Json file should be written following this format:
 - Claim 0: |{| Model Type, General LLM|, |Model Name, ChatGPT-3.5-turbo|, |Parameter Size, 175B|, |Dataset, Spider dev|, |Difficulty Level, 1|}, Execution Match , 0.760|
 - Claim 1: ..
 - Json format containing claims:
 - "[
 - '0': {"specifications": {"0": {"name": "Model type", "value": "General LLM"}, "1": {...}}, "Measure": "Execution Match", "Outcome": "0.760"},
 - '1': {...},
 -]
 - **Notice that specifications are numbered!**
- "paperID" is the id of the paper
- "tableID" is the numbered table (not the actual ID it was reported in the html)
 - First table of the paper has table_id = 1, second table has table_id=2 and so on.
 - If the paper id is "2456.7563" then the file with the associated claims from table 1 is named: "2456.7563_1_claims.json"

Task 1: File and Claims Formats

- Put all claims in a folder called:
 - YOUR_NAME_CLAIMS
- In which **each file** is named
paperID_tableID_claims.json

Task 2: Profiling

- Produce a profiling of the extracted claims.
 - Distributions of “name” in specification.
 - Distributions of “values” for each name of each specification.
 - Distributions of “ metrics”.
- Produce a spreadsheet with ColumnA key and ColumnB number of items.
- Filename should be NAME_PROFILING.CSV (orxlsx)

Task 3: Alignment

- Align specifications names, values and metrics.
- Example:
 - In some experiments, “dataset” might be mentioned as dataset or benchmarks. Or “model” as model or algorithm.
- JSON file for the terms aligned and reproduce the profiling based on these new information.

Task 3: Example of Alignment of Claims

- Claims to align:

1. $\{ | \text{Model Type, General LLM} |, | \text{Model Name, ChatGPT-3.5-turbo} |, | \text{Parameter Size, 175B} |, | \text{Dataset, Spider dev} |, | \text{Difficulty Level, 1} | \}$, Execution Match , 0.760 |
 1. From paperid "1234.5678" table 2
2. $\{ | \text{Model, SMBOP + GRAPPA} |, | \text{Dataset, Spider development set} | \}$, Execution Match , 75.0 |
 1. From paperid "6767.9898" table 4
3. $\{ | \text{Model, Ours (w/ Graphix-T5)} |, < \text{Difficulty, Medium} |, | \text{Dataset, Spider} | \}$, Execution Match , 80.7 |
 1. From paper_id "3859.9017" table 1

Task 3: Example of Alignment of Claims

Model type	Model name	Parameter Size	Dataset	Difficulty	Metric
General LLM	ChatGPT-3.5-turbo	175B	Spider dev	Level 1	Execution Match
-	SMBOP + GRAPPA	-	Spider development set	-	Execution Match
-	Ours (w/ Graphix-T5)	-	Spider	Medium	Execution Match

Task 3: Example of Alignment of Claims

Json File:

```
{
  "aligned_names": {
    "model type": ["1234.567_2_0_0"],
    "model name": ["1234.567_2_0_1", "6767.9898_4_0_0", "3859.9017_1_0_0"],
    ..
  }
  "aligned_values": {
  }
```

- In “aligned_names” and “aligned_values”, for each aligned names and values you have to report as **paperID_tableID_claimID_specificationID**
- You can choose the name you prefer for the aligned value or name.

Task 3: Filename

- Filename for the alignment is
 - YOUR_NAME_ALIGNMENT.JSON

Termini di consegna

- Preparare un documento che describe:
 - La soluzione usata per l'estrazione dei claim
 - Il numero di articoli e tabelle analizzate
 - La soluzione usata per valutare la correttezza dell'estrazione
- Il documento e uno zip contenente i file json per i task descritti sopra vanno consegnati entro il 10 gennaio 2025 attraverso il modulo all'indirizzo:

<https://forms.office.com/e/5nmvtKgY11>