

SemTab 2024
 Semantic Web Challenge on Tabular Data to Knowledge Graph
 Matching

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About the Challenge

Tabular data is the form of data that is the common input format in data analysis pipelines. However, a lack of understanding of the semantic structure and meaning of the content may hinder the data analytics process. Thus gaining this semantic understanding will be very valuable for data integration, data mining, and knowledge discovery. In this challenge, we want to encourage systems to better understand what the data can be used for and how to assess what sorts of transformations are appropriate on the data.

Data on the Web may also be the source of highly valuable data. The addition of semantic information to web data enables a wide range of applications, such as web search, question answering, and Knowledge Base (KB) construction.

Tabular data to Knowledge Graph (KG) matching is the process of assigning semantic tags from KGs to the data. Whereas or Equipped to the task, we invite researchers to participate in the challenge.

The SemTab 2024 challenge aims at benchmarking systems dealing with the tabular data to KG matching problem, so as to facilitate their comparison on the same basis and the reproducibility of the results.

The 2024 edition of this challenge will be collocated with the [21st International Semantic Web Conference](#) and the 19th International Workshop on Ontology Matching.

Challenge Tracks

Accuracy Track Track page →

The main goal of this system regarding accuracy is similar to prior versions of the SemTab. That is, to measure the accuracy of the submissions, we evaluate systems on typical multi-class classification metrics as detailed below. In addition, we adopt the “score” for the CTA tasks to reflect the distance in the semantic space between the predicted column type and the ground truth semantic type.

Matching Tasks

- **CTA Task:** Assigning a semantic type to a database class as fine-grained as possible to a column
- **CER Task:** Matching a cell to a Wikidata entity
- **CSP Task:** Assigning a semantic relationship between two columns
- **RAT Task:** Assigning an KB entity to a table row
- **Table Topic Detector:** Assigning a KG class to a table

STT vs LLMs Track Track page →

This track involves the usage of LLMs to perform the CTA task setup introduced. Participants will be asked to submit their dataset containing semantic annotations. Subsequently, an automatic validation set will be provided to measure the quality of the approaches with the usual metrics used in the previous edition of this challenge. This task considers several challenges, including identifying a way to handle large datasets with an LLM, determining methods to manage memory limits, enriching the training dataset to increase the generalization capability of the LLM, and more, leaving effective prompts to carry out fine-tuning.

Matching Tasks

CTA Task: Matching a cell to a Wikidata entity Track page →

This track invites participants to match table records only, e.g., certain series, to Wikidata entities. This is a challenging task due to the limited available context that could be used by annotation systems to perform the semantic linking. LLMs are a promising way to achieve such tasks that could be utilized in different ways.

TableTopicDetector Task Track page →

Since data quality metrics in this setting are far from being well-defined enough to accurately assess the quality of semantic table annotation detection, manual inspection could be a key solution to answer this question, but what about large-scale datasets? When each of which contains hundreds of thousands of tables? Random-based checks could be an alternative but would it be a good enough solution? What we think is a promising solution is an automated way that runs specific tests on a given dataset. If then

Dataset Revision Contributions

The data that SemTab Knowledge-Graph matching systems are trained and evaluated on, is critical for their accuracy and relevance. We invite dataset submissions that provide challenging and accessible new datasets to advance the state-of-the-art of table-to-KG matching systems.

Preferrably, these datasets should include along with the data, ground truth annotations for at least one of CTA, CER, CSP, RAT, or STT tasks. The submission may either be specific to a certain domain.

Submissions will be evaluated according to provide the following:

- Description of the data collection, curation, and annotation processes
- Availability of documentation with insights in the dataset content
- Details of the data collection, curation, and annotation processes and its DOI
- Explanation of maintenance and long-term availability
- Clear description of the envisioned use-case
- A GitHub repository that can be used to solve an exemplary task

Dataset Revision Contributions Track page →

Revisions entirely new datasets, or also encourage revisions of existing datasets and their annotations. Revisions can be of any kind as before, but we welcome alternative revisions:

- Revised annotations with improved quality
 - New annotations
 - New annotations for an existing dataset enabling new tasks on it
- Please clearly describe and illustrate what the problem is that the revision addresses, and how the addressed approach yields a high-quality dataset for downstream applications. Dataset and annotation revisions are expected to be made public with a permissive license for wider use in the community.

Artifacts Availability Badge

Also this year we have Artifacts Availability Badge which is applicable to the Accuracy Track as well as the Datasets Track.

The goal of this badge is to motivate authors to publish and document their systems, code, and data, so that others can use them and build upon them. It is a great way to share your work!

This badge is given if all resources are verified to satisfy the below criteria. The criteria used to assess submissions (both accuracy and dataset submissions) are:

- Publicly available [GitHub]
- Publicly accessible source code
- Clear documentation of the code and data
- Open-source dependencies



Paper Guidelines

We invite papers to submit using [arXiv](#).

System papers (including Accuracy, STT, LLMs, Only Metadata to KG, and Datasets Tracks) should be no more than 12 pages long (excluding references and figures for the Datasets Track are limited to 8 pages). If you are submitting to the Datasets Track, please append “[Datasets Track]” at the end of the paper title.

The papers should be formatted using the [CLUEW Latex template](#) or the [CLUEW Word template](#). Papers will be reviewed by I-2 challenge organizers.

Accepted papers will be published as a volume of [CLUEW](#). By submitting a paper, the authors accept the CLUEW publishing rules.

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