Intuitive Semantic Graph Tool for Enhanced Archive Exploration

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The work

To define a pipeline able to create semantic graphs in an Intangible Cultural heritage archive

with two aims:

1. To offer a global view of the content of the archives for any users;
2. To enhance traditional ways of searching and browsing data on the web.

Innovative way to visualize the contents of an archive, as a multilevel graph

Characteristics of the approach

- The pipeline is completely **unsupervised**
- Semantic graphs are **layered**, in case of very large archives
- Use of **pretrained language** models makes it suitable also for a few hundred items
- **Multilanguage** o language specific pretrained language models make it suitable for documents in English, Italian, French, even mixed.

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Innovative elements of pipeline

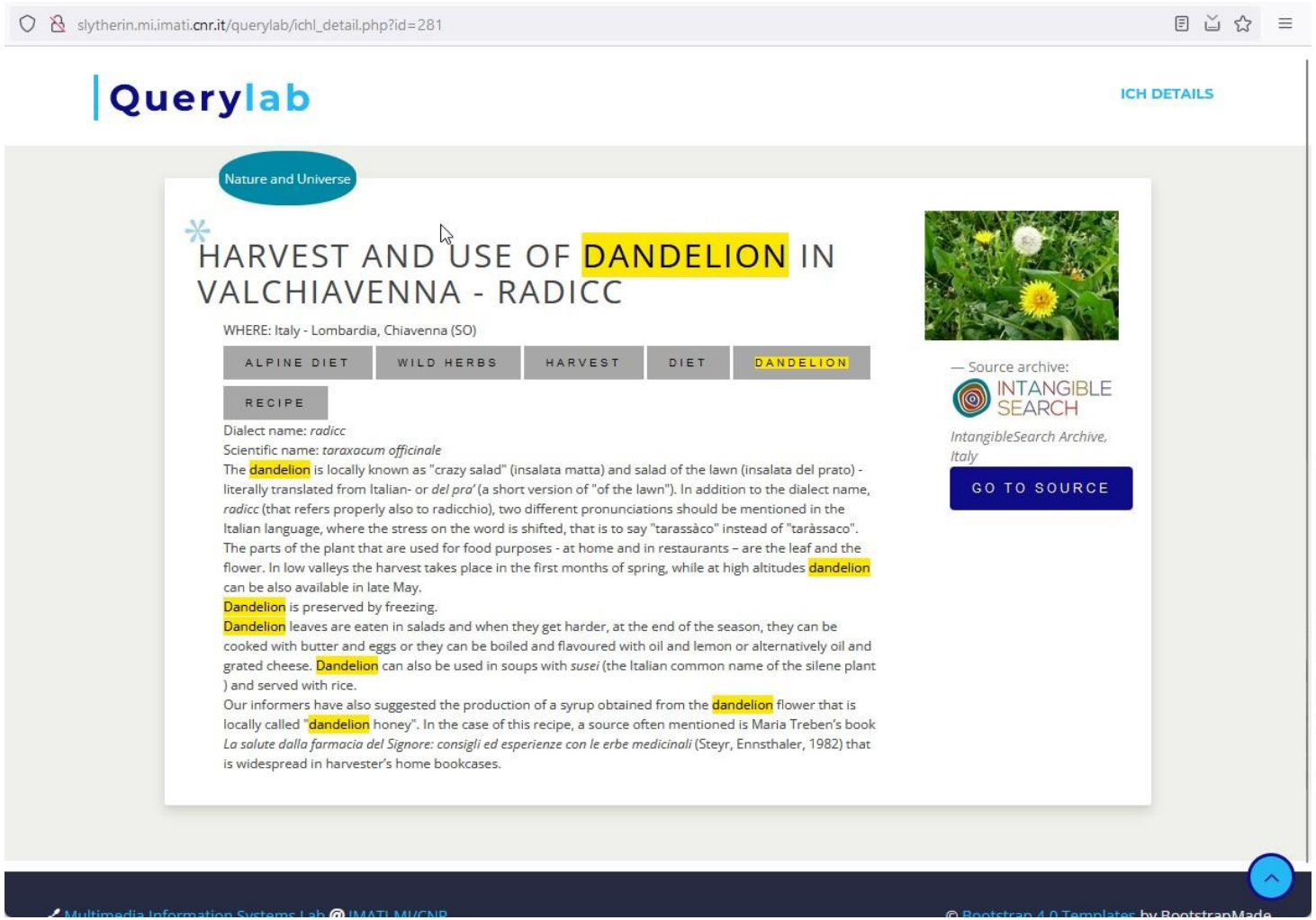
1. **UMAP** is a dimensionality reduction technique. UMAP uses manifold learning for mapping high-dimensional data to a lower-dimensional space while preserving the local structure of the data
2. **HDBSCAN** is a clustering algorithm that groups similar data points together. HDBSCAN does not require the user to specify the number of clusters or the size of the neighbourhood to be searched. It automatically detects the number of clusters and the shape of clusters
3. **Transformers** are a type of neural network used in natural language processing. Transformers use self-attention mechanisms to focus on different parts of the input and capture long-term dependencies.

The data: Querylab

A portal specifically designed to manage intangible cultural heritage data.

Two types of data:

- Data stored locally
- and data queried on the fly from remote repositories via REST API web services.



The data: Querylab

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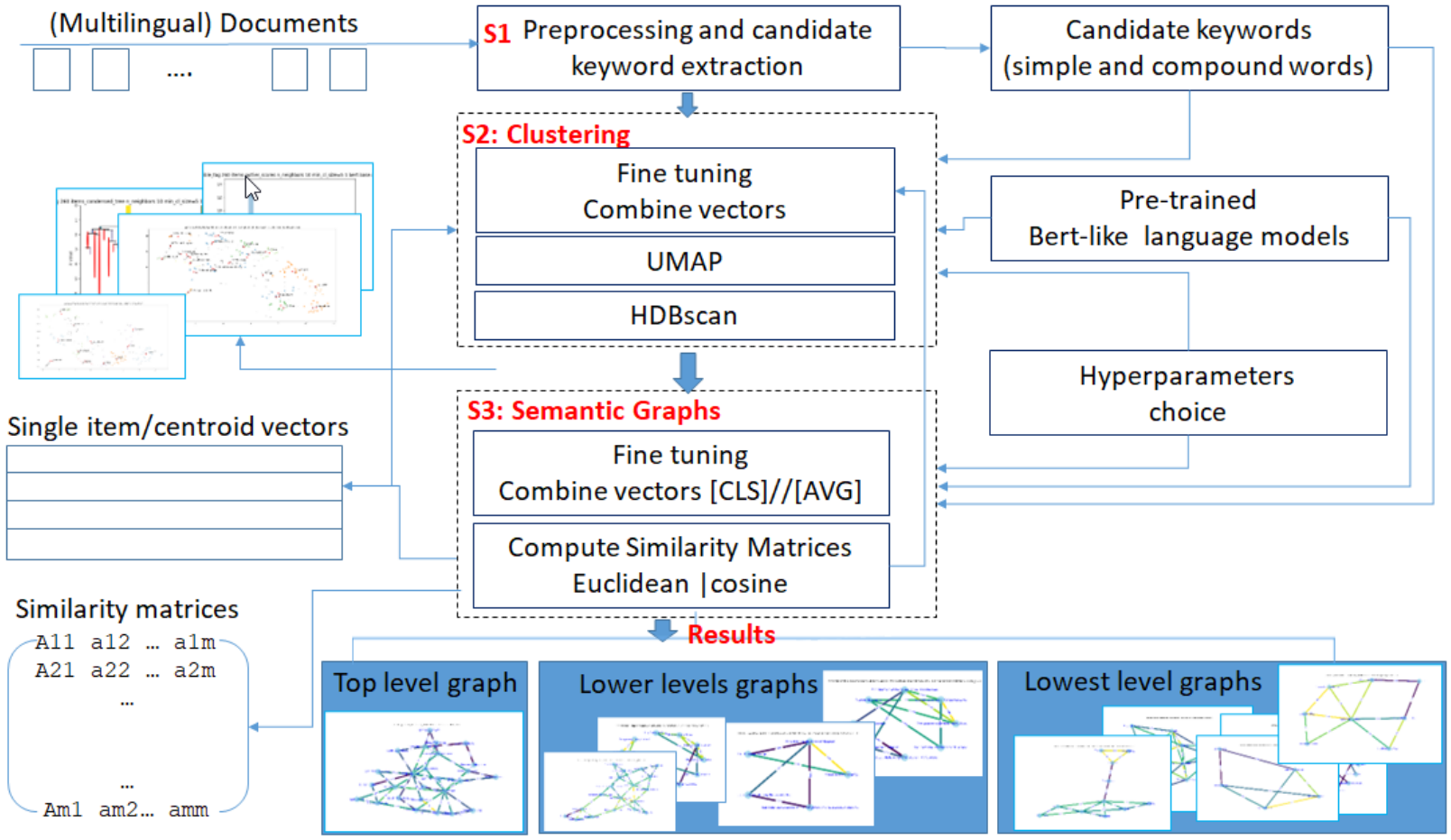
The data: Querylab - 2

1. **Tags**: Expert-defined tags associated with the records in the archive
2. **Title**
3. **Description**
4. **Rake/Textrank Keywords**: simple or compound words were in an automatic and unsupervised manner extracted from descriptions



The proposed pipeline

 # Task 1: Dataset Preparation

- Preprocessing (possibly strip stopwords, accents, …)
- Process data to extract items to be used
- Output: items of interest

 # Task 2: Items clustering

- Choice of transformers and pre-trained models
- Fine tuning of pre-trained Bert-like models to obtain the vectors
- Choice of hyperparameters for UMAP and HDBSCAN
- Output: centroids of clustered items, and elements of each cluster

 # Task 3 Semantic graph creation

- Choice of transformers and pre-trained models, both on raw data and on clustered items and fine tuning
- Creation of similarity matrix using [AVG] or [CLS] tokens
- Output: Semantic graphs con k most similar items, with k=1…4
- Preliminary evaluation of the results with domain experts and web users

The proposed pipeline

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Dataset preparation

1. preprocess
2. extract items of interest

Items of interest: either short texts or set of terms, single or compound words

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Items clustering

- Performed using **UMAP and HDBSCAN** on the vectors obtained by tokening items of interest.

  - Choice of hyperparameters for UMAP and HDBSCAN

  - n_neighbor (UMAP) : 20,15,10,5.
  - min_cluster_size (HDBSCAN) 15,10,5,
  - min_samples (HDBSCAN) 15,10,5,1

N neighbors	min_cluster_size	min_samples	Number cluster
20	15	5	11
20	15	1	12
20	10	5	14
20	10	1	15
20	5	5	17
20	5	1	30
15	10	1	18
15	5	1	34
10	10	5	13
10	10	1	20
10	5	5	18
10	5	1	33
5	10	1	21
5	5	5	26
5	5	1	45

Items clustering – 2

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Items clustering – 3

- Performed using UMAP and DBSCAN on the **vectors obtained by tokening items of interest.**
  - Choice of transformers and pre-trained languages among these:
  - **BERT Base**: This is the original pre-trained BERT model released by Google. It has 12 transformer layers and is trained on a large corpus of text data from Wikipedia and the Book Corpus dataset.
  - **DistilBERT**: a distilled version of BERT model: smaller, faster, cheaper and lighter.
  - **MiniLM-L6-v2**: This is a smaller version of the BERT model developed by Microsoft. It has only 6 transformer layers and is trained on a subset of the data used to train BERT Base.
  - **Bert-base-Wikipedia-sections-mean-tokens**: This is a pre-trained BERT model released by the Hugging Face team. It is trained on a large corpus of text data from Wikipedia and uses a mean pooling strategy to create a fixed-length representation of the input text.

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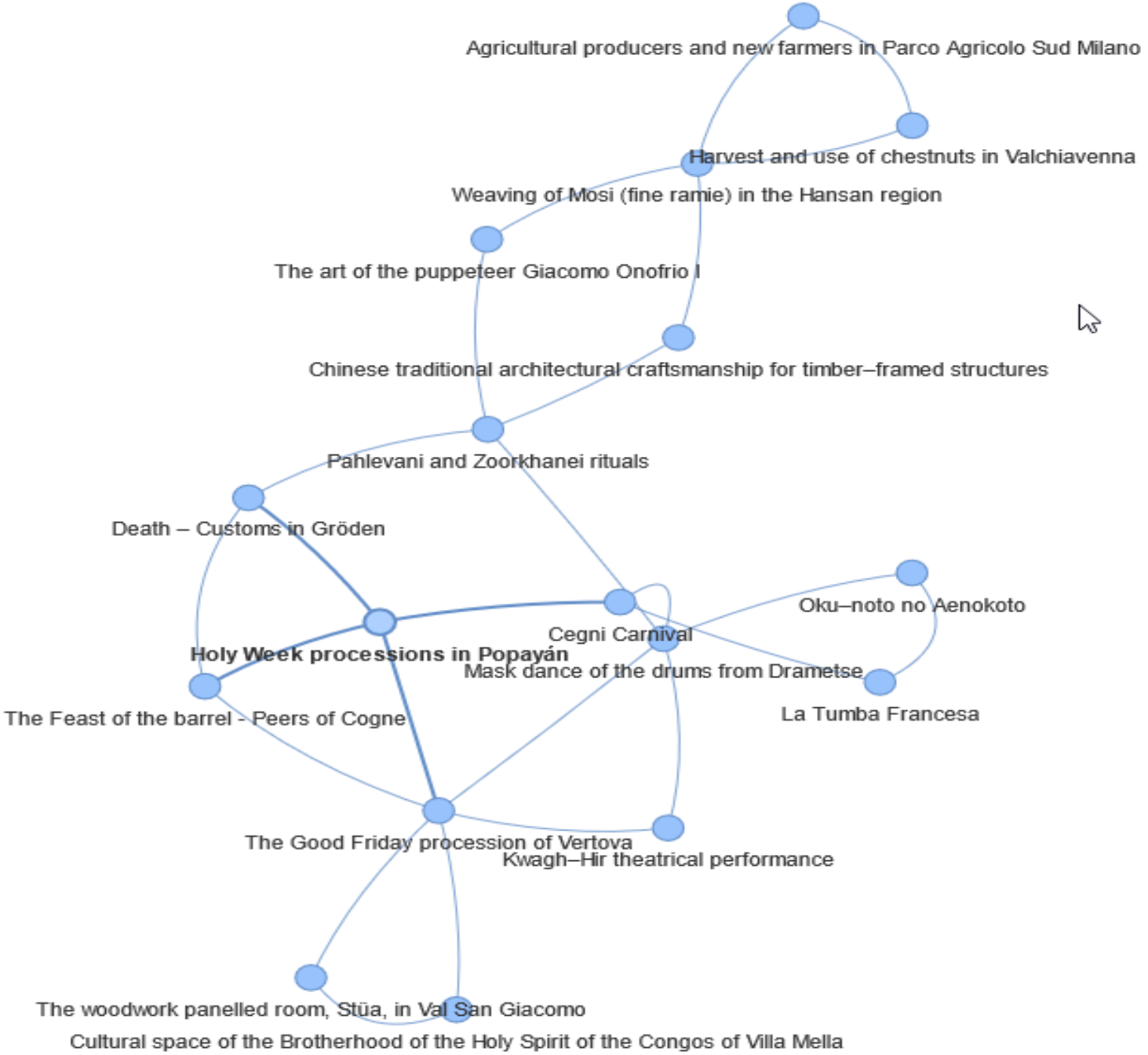
Semantic Graph Creation

- Creation of similarity matrices, for the centroids and the clusters
- Use of [CLS] or [AVG] to create a single vector per item (title, tag, description, …)
- K-most similar items, with k ranging from 1 to 4

Whole dataset

With k=2

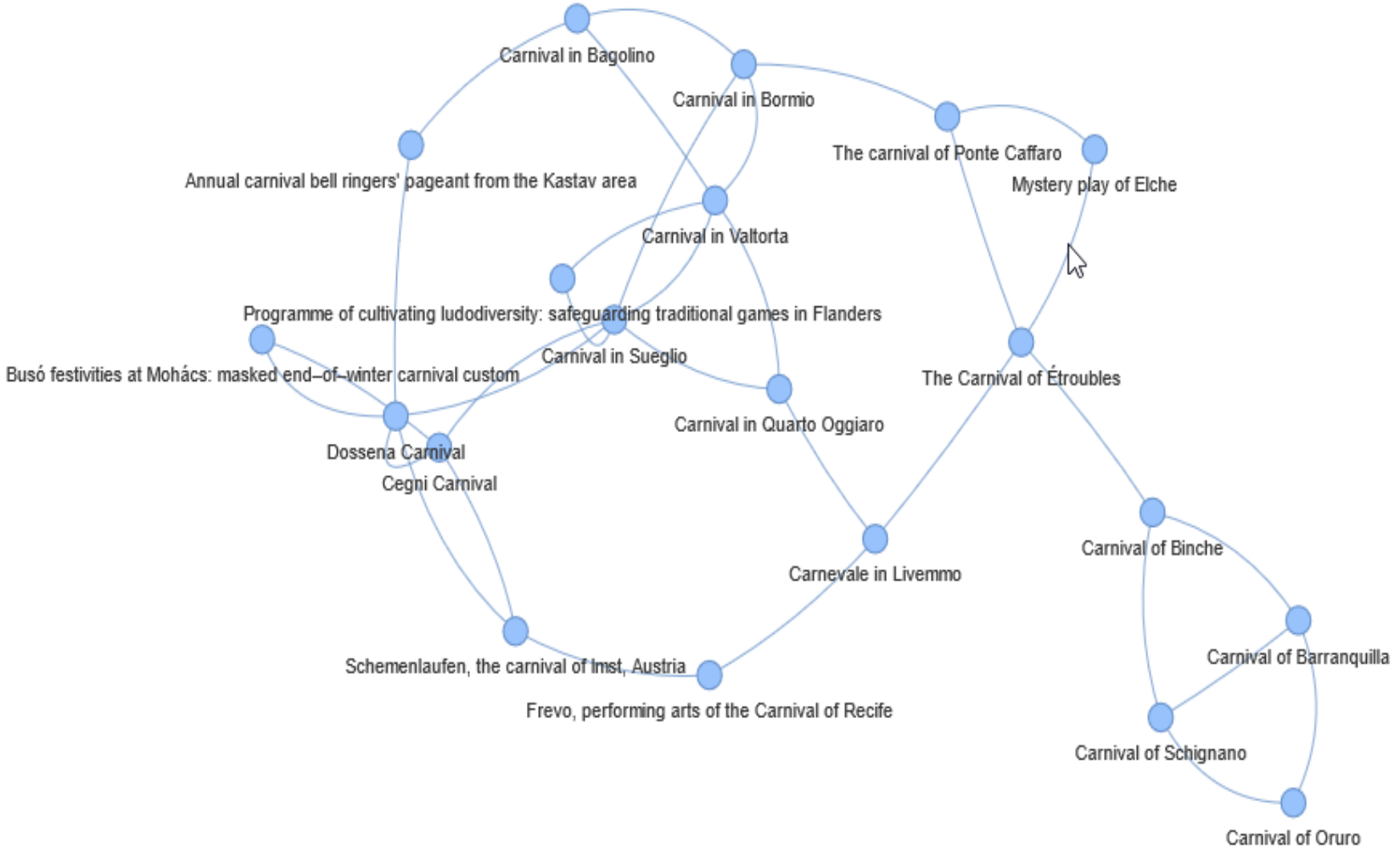


Cegni Carnival

With k=1

Cegni Carnival

With k=2



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Evaluation

Qualitative evaluation

We gathered feedback from heritage experts and web users:

- is the clustering and similarity matrix able to extract the significant elements?, and
- Do users find browsing the archive via graph interesting and useful?

- **Positive results:**
  - **PROs**: simplicity and usability of the graph visualization
  - **CONs**: low-level clusters contained elements that were not closely related, or that some related elements were spread across multiple clusters

Conclusion

Definition of a pipeline for the creation of semantic graphs as a layered map with different granularity

New way of searching and browsing ICH archives

Preliminary evaluation

- Effectiveness of the pipeline in generating meaningful semantic graphs
- Positive evaluation from users, but graphs with more than 30 nodes (overly dense graphs) are difficult to understand and navigate

Future works

- tools to traverse the graphs
- fish-eye views to overcome the overly dense graphs
- experiments on other datasets

Thank you!

For any question

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