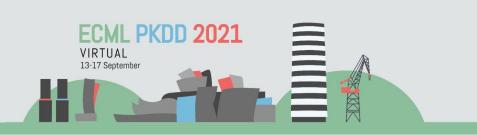
Web Image Context Extraction with Graph Neural Networks and Sentence Embeddings on the DOM tree

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This paper introduces a novel approach for Web Image Context Extraction (WICE) that combines Graph Neural Networks (GNNs) and Natural Language Processing models.

Introduction

- ❖ Identifying the text in a webpage that best describes an image is a key step for efficiently indexing images in a search engine
- Visually rendering the webpage facilitates the extraction of an image's context, but isn't tractable on a large scale

Our Contribution

Use state-of-the-art language models to generate sentence embeddings for each text node in the DOM tree



Fig 1: An example of the WICE setting

- Use sentence embeddings as node features to train a GNN, which can combine both structural and semantic information
- Use graphe models for large-scale processing of highly diverse news websites

Use HTML document to predict reference text embedding (at training time) Reference Text Embedding

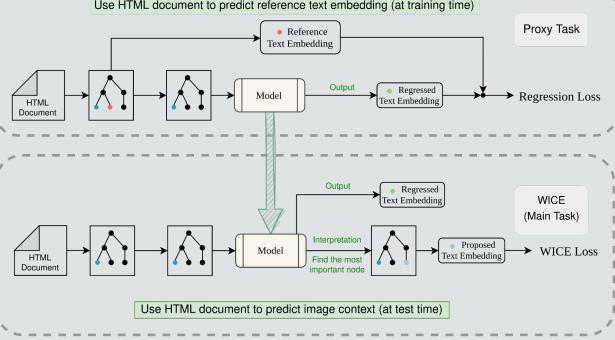


Fig 2: Our pipeline

- Proxy task: train a GNN model to predict the input document's reference text (red dot)
- Main task: interpret the trained model to choose the most predominant textual node (green dot) is then used as the context of the image

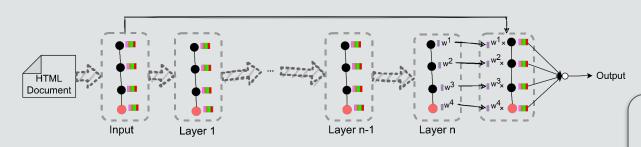


Fig 3: weight-GCN model architecture

Experiments

| Method | train | validation | test |
|------------------|-------|------------|-------|
| random | 0.78 | 0.779 | 0.779 |
| title | 0.834 | 0.835 | 0.833 |
| text after image | 0.671 | 0.672 | 0.67 |
| wGCN | 0.381 | 0.386 | 0.381 |
| oracle | 0.293 | 0.297 | 0.293 |

Table1: average cosine similarity loss between the context and the reference text [split dataset by webpages]

| Method | train | validation | test |
|------------------|-------|------------|-------|
| random | 0.792 | 0.736 | 0.800 |
| title | 0.834 | 0.861 | 0.814 |
| text after image | 0.701 | 0.571 | 0.705 |
| wGCN | 0.415 | 0.404 | 0.441 |
| oracle | 0.334 | 0.264 | 0.259 |

Table2: average cosine similarity loss between the context and the reference text [split dataset by websites]

Our model significantly outperforms WICE based on heuristics, and can work directly with HTML, making large-scale WICE tractable.