A model-based framework for NEWS content analysis

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- Keywords: Category theory, Content analysis, Model based framework, Knowledge graph, Natural language processing, Computational journalism.
- News articles are published all over the world to cover important events. Journalists need to keep track of Abstract: ongoing events in a fair and accountable manner and analyze them for newsworthiness. It requires enormous amount of time for journalists to process information coming from main stream news media, social media from all over the world as well as policy and law circulated by governments and international organizations. News articles published by different news providers may consist of subjectivity of the reporters due to the influence of reporters' backgrounds, world views and opinions. In today's practice of journalism there is a lack of computational methods to support journalists to investigate fairness and monitor and analyze large massive information streams. In this paper we present a model based approach to analyze the perspectives of news publishers and monitor the progression of news events from various perspective. The domain concepts in the news domain such as the news events and their contextual information is represented across various dimensions in a knowledge graph. We presented a multi dimensional comparative analysis method of news events for analyzing news article variants and for uncovering underlying storylines. To show the applicability of the proposed method in real life, we demonstrated a running example in this paper. The utilization of a model-based approach ensures the adaptability of our proposed method for representing a wide array of domain concepts within the news domain.

1 Introduction

In every human community in the world human beings bring news to one another. It has an important role in humankind and journalists are involved in carrying out the task in a professional way. While reporting about a real life events into news articles, journalists turn facts into stories and analysis that engage an audience (Schudson, 2020). While making news, good journalists put reality first and they follow the core principles of ethical journalism (EJN, 2023) which include *trust and accuracy*; *independence*; *fairness and impartiality*; *humanity*; *accountability*. However there is no bias-free journalism (Schudson, 2020) in reality. The problem of bias in media has been an important topic and it requires sophisticated techniques to analyze the media bias in a

systematic way. Journalists also need to keep track of the ongoing events all around the world and analyze the events carefully as they need to inform their audience about the changing world. Since there is an abundance of news articles being published all over the world by several news media outlets, journalists would be benefited by techniques for systematically analyzing events from news publications. Sociologists, historians, political scientists, information scientists are involved in gathering information from news articles and extract insightful information. In this paper we present a model based framework that employs a diverse range of models for representing the knowledge in the domain and for representing computational methods for the analysis of news events. The framework integrates the following components:

- state-of-the-art natural language processing technique for parsing content from news articles;
- a dimensional meta-model allowing data to be arranged into hierarchical groups and a knowledge

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graph schema for structuring event related information;

- a content comparison method using category theory approach;
- a statistical analysis method for analyzing news article variants;

The knowledge graph represents news events with relevant information e.g., source article, publication date, involved persons, involved countries, and type of event. We annotate news events with IPTC (International Press Telecommunications Council) Media Topics. IPTC is a global standards organization that provides metadata standards for the news industry. The terms in the IPTC Media Topics are represented in a hierarchical structure which allows us to extract news events from different abstraction levels. By combining different attributes and relationships of news events along with the domain ontology in IPTC Media Topics, the framework allows users to extract different views of news events from a knowledge graph. The framework integrates a computational model based on category theory which allows us to analyze news events at a higher abstraction level, for example, to compare and categorize events; to analyze flow of progression of events. We present novel application areas of category theory for analyzing events stored in a knowledge graph. We assume the reader is comfortable with the basics of category theory (Barr and Wells, 1990).

In section 2 we present a method of extracting structured information about news events from news articles using large language models. We present a running example while describing the proposed method. In section 3 we present a model-based framework for content analysis. In section 4 we provide a discussion about the proposed method and provide a comparison with existing works.

2 Harvesting News Events Knowledge Graph with a Pre-Trained LLM

Harvesting news events into a knowledge graph is an important topic and it has been addressed by several other projects to perform various tasks in the news domain. Opdahl et. al. (Opdahl et al., 2022) provides a review of using semantic knowledge graphs in news production, distribution, and consumption, emphasizing their potential for integrating heterogeneous information in the news industry. The Global Database of Events, Language, and Tone (GDELT) is a project sponsored by Google that monitors news media from all over the world and provides a real-time update of events every 15 minutes (Gde, 2023).

Rospochera et. al., present a method to automatically build Event-Centric Knowledge Graphs from news articles using NLP techniques, such as Entity linking and Semantic Role Labeling (Rospocher et al., 2016). Liu et. al., introduces a domain-specific knowledge graph called the "news graph" that incorporates collaborative relations between entities and topic context information for news recommendations (Liu et al., 2019).

Harvesting news events into a knowledge graph has been studied by Berven et. al., in (Berven et al., 2020) where they presented a knowledge graph platform for newsroom. They propose an event detection technique that identifies potentially newsworthy events from clusters of news items according to named entities, topics, and location.

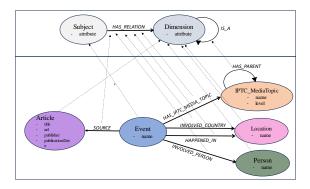


Figure 1: Dimensional meta-model (top) and Knowledge Graph Schema (bottom) for structuring event related information

In our proposed technique, we take input from GDELT every 15 minutes. The input includes web addresses to news article texts. The news article texts are parsed for analysis using a pre-trained large language model. We use GPT-Turbo 3.5 for extracting information from news article text and harvest news event related information. To structure the news events information we propose to use a dimensional metamodel (Figure 1 top) which allows storing events with contexts along various dimensions in a hierarchical model. The bottom of Figure 1 shows a knowledge graph schema for structuring an event and its contextual information such as event location, event type and involved countries. The knowledge graph is also enriched with IPTC Media Topics. The knowledge graph allows us to access the hierarchical information from the IPTC Media Topics ontology by traversing over :HAS_PARENT relationships. A Neo4j graph database has been used to store news events and their

relationships with other entities. The information model is centered around *Event* and it also allows us to preserve the epistemic view of individual publishers. For example, if two publishers publish 2 news articles about a certain event, we will be storing 2 instances of *Event* (along with their contextual information) in our knowledge graph.

Table 1 illustrate the prompt we have used to extract structured information from news article texts. The temperature is set to 0 to limit the creativity of the large language models. The prompt includes instruction about producing output in JSON format. It also includes instructions to classify events using IPTC Media topic names but the GPT 3.5 Turbo model generates slightly different names from what we have in the knowledge graph. For example, in our knowledge graph we have 'arts, culture and entertainment' but the output of the prompt may include the following name: 'Cultural, Arts and Entertainment'. It is therefore necessary to perform a similarity analysis of the media topic names. To find out the most similar topic name we have calculated cosine-similarity using Spacy's Python library.

Table 1: Prompt for extracting event related information

prompt = "Write the name of the event, type of the event, involved person, involved countries and the location of the event from the following news. Use IPTC media topic name while writing values for 'Event type'. Write full name while mentioning involved persons and locations. Write only name of persons if they are known. No need to include any unknown person. Also do not need to write the designation or position of the persons. While returning the location, mention the country where the event took place. While returning the iptc media topic names, please return the output for which you are significantly confident about. If there are more values, include all of them in comma seperated format. Format your answer as a JSON object with the following key-values: {"Event': "event-name' "Event Type': "iptc-media-topic-name'', "Involved Countries': "country-name'', "Location of Event': "country-name'', "Involved-Person ': "Person-name ', }" response = openai.ChatCompletion.create(model="gpt-3.5-turbo", messages=["role": "system", "content": prompt, "role": "user", "content": articleText], temperature=0, max_tokens=256, top_p=1, frequency_penalty=0, presence_penalty=0) 'Event': "Closure of Niger's Airspace' 'Event Type': 'Civil Unrest', 'Involved Countries': 'Niger, United Kingdom, South Africa', 'Location of Event': 'Niger', 'Involved-Person': 'President Mohamed Bazoum, General Abdourahmane Tchiani' }

The proposed method in this paper is demonstrated with a running example which includes a knowledge graph of news events about *Niger* and *Gabon* extracted from the news articles published by 6 media outlets (*aljazeera.com, theguardian.com, reuters.com, independent.co.uk, nytimes.com, washingtontimes.com*) from 28. July, 2023 to 02. Sept 2023. The knowledge graph therefore consists of news events in *Niger* and *Gabon* about two coup's which took place during the above mentioned period.

3 Model-based Framework for content analysis

We propose a new model-based framework for news content analysis that includes techniques for multi dimensional comparative analysis. The framework allows us to analyze the perspectives of news contents; progression of events from a variety of abstraction levels with various perspectives. The framework allows the user to select an appropriate dimension and abstraction level. For instance, a user might be interested in comparing the perspectives of different publishers over a certain period of time or the progression of events at a certain level of abstraction. The knowledge graph includes events and its contextual information at various dimensions represented in hierarchical groups, for example, the IPTC Media topic ontology includes information at different hierarchical groups. The highest level of abstraction in the IPTC Media topic ontology (i.e., level 1 of the ontology) includes 17 media topic names. The selection of the dimension and abstraction level is used for extracting information from the knowledge graph. This information from the knowledge graph is used for a comparative analysis. The analysis results are used for extracting patterns of variants. We propose a semi-automated approach where humans are involved in the process of variant analysis. Figure 2 illustrate a model-based framework which employs models for representing computational methods for the analysis of news events. Graph patterns are used to specify search criteria. We propose to use categorial operations to perform comparative analysis over the search results (i.e., subgraphs). Category theory allows us to deal with abstract structures and relationships between them. It allows us to study the news content from high level of abstraction and therefore enables us to gain deeper insights into media contents. In this paper we focus on the analysis of perspective comparison, progression of events, and variant analysis. The model-based framework is adaptive to new dimensions with more contextual information, for example,

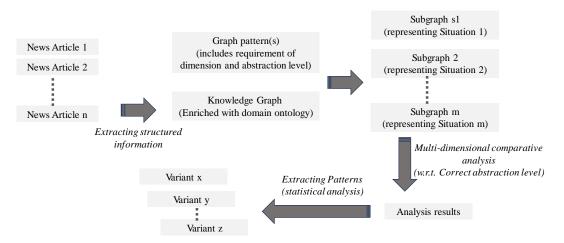


Figure 2: Model-based framework for multi dimensional comparative analysis of news contents

number of casualty, sentiments, proximity, news angles, etc.

3.1 Analysis of perspective comparison

We store news events contextual information such as event location, event type, involved countries, and individuals in the knowledge graph. When reporting on specific events and their subsequent developments, various publishers may have reported them differently. In our proposed method, we compare the perspectives across various dimensions of these events. For instance, we examine the types of events that were reported by different publishers during a specific time period while they were covering a particular event and its subsequent development.

To compare the perspectives of different publishers, we propose employing category theory operations, including pullback and commutative diagrams. Figure 3 gives an overview of the proposed method for perspective analysis. All the news article related information in the graph database is represented as Iin the figure. O_1^* and O_2^* represent the reports from two different publishers. O_1^* and O_2^* can be computed by querying the graph database. Cypher queries (Cyp, 2023) may be used to extract the fragments of graphs (i.e., subgraphs) from I which represents the local perspective of individual publishers. For instance, we may be interested to know the extent to which the media topics used by different publishers match and differ while they report about some events in their published news articles. The figure shows pullback object C which is computed from the following two morphisms: $O_1^* \xrightarrow{m_1} I$ and $O_2^* \xrightarrow{m_2} I$. From the pullback object, we can figure out the perspectives of different publishers as shown in Figure 3 by object D_1 and D_2 .

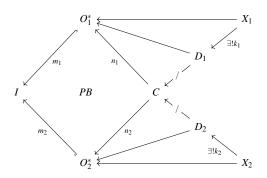


Figure 3: Pullback object (C) computes the commonality between O_1^* and O_2^* ; D_1 and D_2 objects are used to compute the dissimilarities between O_1^* and O_2^*

Here we explain the proposed method with a running example. We will compute the perspectives of two publishers about their news stories that covers the events in Niger from 28 - July to 02 - September. We use Cypher queries to get data from the Neo4j graph database. Cypher queries can be expressed as graph patterns which includes variables. The results of these queries would be subgraphs of the whole graph database. One can compute the pullback object by writing a program using general purpose programming language (e.g., Python using the Neo4j library) but in this paper we present a Cypher query (Figure 4) which computes the pullback object of two subgraphs from a graph database by combining two Cypher queries as shown above. We ensure that the diagram commutes by specifying t1 = t2 as a condition in the query. Since the two subgraphs s_1 and s_2 include only nodes of type IPTC_MediaTopic, we include IPTC_MediaTopic nodes in the result pullback object. Figure 4 shows a cypher query expression to compute the pullback object of $s_1 \rightarrow I$ and $s_2 \rightarrow I$.

The perspectives of the publishers are computed

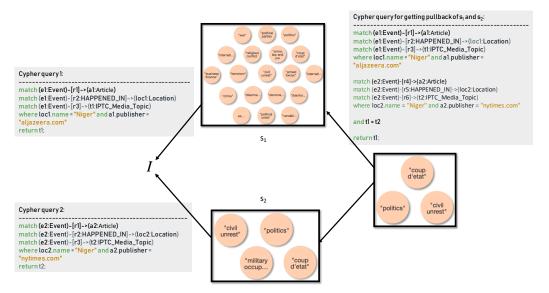


Figure 4: Computing pullback with Cypher query

from the difference of the subgraphs S_1 and S_2 with the pullback object. Here we have demonstrated the perspective analysis with respect to IPTC media topics but the other dimensions can also be used for perspective analysis.

3.2 Analyzing the progression of events

To analyze the progression of events in computational journalism is a complex task as there is an abundance of information. Numerous publishers from all around the world publish about ongoing events. There is a lack of tool support in computational journalism to keep record of the events and systematically analyze them for extracting insightful information about the progression of events. We propose (1) to use features such as names, locations and IPTC topics to group news articles covering stories about closely related topics and, then, (2) to use a categorical approach to analyze the progression of events by means of analyzing contents in news articles. We reuse the concept presented in Figure 3 where we adapt O_1^* and O_2^* with a selection of events capturing situations from $time_{x1} - time_{y1}$ and $time_{x2} - time_{y2}$ respectively. From O_1^* and O_2^* we systematically compare the evolution of events from $time_{x1} - time_{y1}$ to $time_{x2} - time_{y2}$. For example, O_1^* and O_2^* may represent the IPTC media topics being used to cover the news events about Niger from 31.July, 2023 -06.Aug, 2023 and 07.Aug, 2023 - 13.Aug, 2023 respectively. From these subgraphs we compute the emerging IPTC media topics in the reports published during 07.Aug, 2023 - 13.Aug, 2023. This comparative analysis allows journalists get an overview of the progression of events.

The progression of events can be represented as a transformation of IPTC media topics being covered by the publishers. Let us consider that in Figure 5 Niger_{28.07-30.08}, Niger_{31.08-06.08} and Niger_{07.08-13.08} are representing the IPTC media topics being used to cover the news events in Niger for period 28.07-30.07, 31.07-06.08 and 07.08-13.08 respectively. The pullback object C_1 and C_2 represents the commonality of the events (w.r.t IPTC media topics) in Niger_{28.07-30.08}, Niger_{31.08-06.08} and *Niger*_{31.08-06.08}, *Niger*_{07.08-13.08} respectively. The object D_1 would capture the media topics being removed from the reporting during 30.07-06.08; D_2 would capture the media topics being newly added during 30.07-06.08. Similarly, D_3 would capture the media topics being removed from the reporting during 07.08-13.08 and D_4 would capture the media topics being added during 07.08-13.08.

Similar categorical operations can also be used to analyze the progression of events in two different countries. Let us consider that we want to analyze the weekly progression of events in *Niger* and *Gabon* since the coup started in those two countries. Figure 6 illustrates a computational model for such analysis. The pullback object C_{AB1} is the commonality between the progression of events in the two countries *Country* – *A* and *Country* – *B* in the first week where *Country* – A_{w1} and *Country* – B_{w1} represents contextual information of events (such as IPTC media topics or involved countries or individuals) reported in the first week; For brevity we did not show

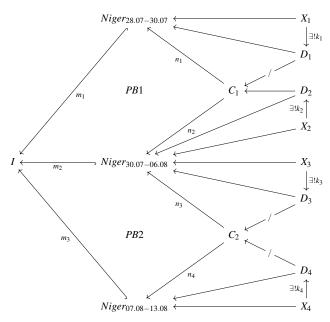


Figure 5: Capturing the progression of events with pullback operation

 C_{AB2} (pullback object between $Country - A_{w2}$ and $Country - B_{w2}$) in the diagram. Common development between the two countries progression can be found from the pullback objects C_{AB1} , C_{AB2} , C_{AB3} , etc.

Figure 7 illustrates a computation model for the comparison of progression of events at a higher level of abstraction. $\alpha_1, \alpha_2, \beta_1, \beta_2$ represents contextual information of events specified at a certain abstraction level *j*; In our running example we only have hierarchical data model for IPTC Media topics, therefore, all the IPTC Media topics in $\alpha_1, \alpha_2, \beta_1, \beta_2$ are at level *j* in the IPTC Media topic ontology. $\alpha'_1, \alpha'_2, \beta 1', \beta 2'$ represents contextual information of events specified at a higher level of abstraction. The pullback objects $C_{\alpha\beta i}$ (at bottom layer) represents the commonality of the progression of events. The arrows between layers represent graph homomorphisms between corresponding elements from lower level of abstraction to higher level of abstraction in the KG *I*.

Theorem: For any non-empty pullback object $C_{\alpha\beta i}$ at level *j*, the corresponding pullback objects $C'_{\alpha\beta i}$ at level k < j is non-empty.

Proof sketch: Consider a non-empty pullback object $C_{\alpha\beta i}$ at level *j*; this would require at least one element $n_a \in \alpha_i$ and one element $n_b \in \beta_i$ where n_a and n_b are mapped to the same element in the knowledge graph. If n'_a (with level *k*) is a parent of n_a , and n'_b (with level *k*) is a parent of n_b , then n'_a and n'_b must also map to the same element in the knowledge graph. The pullback objects $C'_{\alpha\beta i}$ should at least contain an

element that maps to n'_a and n'_b and therefore cannot be empty.

3.3 Variant analysis

In this section we present a technique for variant analysis over the results of computation from section.3.2. We present an application of statistical analysis method for detecting news article variants. In section.3.2 we presented techniques to retrieve data from a knowledge graph across various dimensions and based on various abstraction level. This selection of data from knowledge graphs are used for identifying variants by applying statistical methods. In this section we present *Exploratory data analysis* for identifying trends in time and space and use them for variant analysis.

In order to identify trends in reporting across different topics, we need to select a dimension and abstraction level and extract data from the knowledge graph. Suppose we would like to identify trends of publishers reporting about *civil unrest* in *Niger* from 01.08, 2023 – 20.08, 2023, we retrieve the events from the knowledge graph that matches with the *civil unrest* IPTC media topic. The results are therefore used for statistical analysis e.g., frequency distribution and visualization of trends in a timeline. Visualizing the events in timeline would allow us to depict picture about certain type of events being reported by different publishers and their engagement in reporting throughout a selected period of time.

Figure 8 highlights the duration of engagement of individual publishers among *al jazeera.com*, theguardian.com, reuters.com, independent.co.uk, nytimes.com, washingtontimes.com, cnn.com for their reporting about civil unrest in Niger. The background in the figure indicates the co - limit of all the events from these publishers about civil unrest in Niger. From the figure we can extract variants e.g., independent.co.uk and washingtontime.com's similarity during the time of publishing about civil unrest in Niger. However, one might be interested to explore the dataset for identifying trends in other dimensions e.g., the involvement of certain countries in a conflict. Such requirements can be adapted by the proposed method as we can retrieve events those are about any kind of conflict and the events are involving certain countries. We exploit the use of ontological hierarchies for the retrieval of events at the correct abstraction level. For example, we can identify if there are any common trends in the involvement of foreign countries across all the coups that have occurred in African countries.

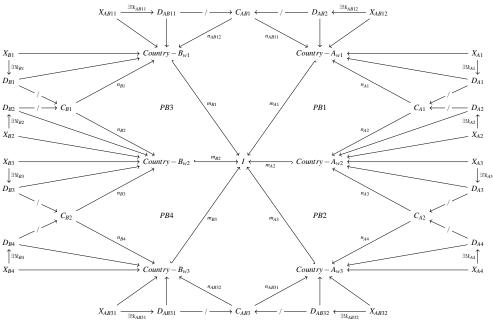


Figure 6: Comparison of progression of events

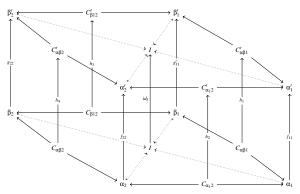
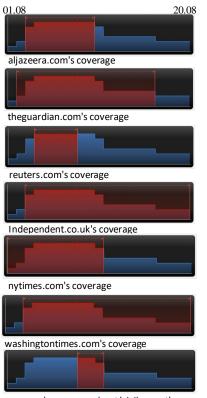


Figure 7: Comparison of progression of events at a higher level of abstraction

4 Discussion and Future work

The proposed method allows us to analyze the perspective of publishers across different dimensions and abstraction levels and we have presented how perspective of publishers covering the types of events can be captured. However, there are many other aspects that might be important to capture such as presentations, opinions, etc. In the landscape of news content analysis, various systems such as GDELT(Leetaru and Schrodt, 2013) have been developed for identifying and organizing news events from vast data streams in structured formats. While GDELT efficiently aggregates and quantitatively analyzes vast volumes of news data, offering an overview of the dynamics within the media landscape, a new approach



cnn.com's coverage about 'civil unrest' Figure 8: Timeframe showing the engagement of news publishers in reporting about *civil unrest* in *Niger*

is needed to enable researchers to dive deeper into individual news events. We presented a model-based framework for content analysis that deviates from traditional news analysis methods that mostly rely on text mining and semantic technologies (Leban et al., 2014; Rudnik et al., 2019). Our proposed method introduces a comprehensive framework that holds the potential to address critical challenges within the media domain. One limitation in the previous research lies in the limited ability to effectively compare news items with one another. Our model fills this gap by offering a robust mechanism for comparative analysis. As a result, our model empowers users to explore and solve open problems in the field of media with a holistic approach, leading to enhanced insights and deeper understanding of the complex media landscape.

In this paper, our primary focus has been on the analysis of various reports pertaining to a specific event, particularly in terms of perspectives. By focusing into the perspective of reports, we aim to uncover the nuances encapsulated within the media discourse surrounding the event. We can furthermore include the intricacies of reporting angles, tones, and the framing of articles, enriching our understanding of news narratives. Additionally, we have employed a systematic approach to track the evolution and progression of these events over time which provides valuable insights into how events unfold and transform over time, enriching our understanding of their dynamics and implications.

Large language models have demonstrated exceptional performance in specific language-related tasks. However, they also fall short in delivering the structured approach and transparency necessary for conducting in-depth multi-dimensional analyses. Our proposed framework, on the other hand, provides a holistic structure for exploring news, ensuring transparency and facilitating a deeper understanding of news content from various dimensions and abstractions. Moreover, our approach distinguishes itself by offering a high level of abstraction combined with the flexibility for users to select different dimensions for exploration. In contrast to LLMs, our approach goes beyond natural language understanding to incorporate statistical analysis, enriching our capacity to uncover nuanced patterns and insights in news content.

While we have presented some analysis technique using category theory, there are much more to explore and develop in this field. We believe that the integration of generative AI and category theory can contribute to the evolution of journalism in the digital age, fostering transparency, accountability, and enriched news content for both journalists and readers. Particularly, our approach has the capacity to assist in tasks that involve the comparison of news items. For instance, it can be particularly useful in Multilingual News Comparison, where it can facilitate crosscultural analysis of news events by overcoming language barriers. Moreover, our model can play a valuable role in Fact-Checking and Verification, aiding in the assessment of news source credibility. Additionally, it is well-suited for Bias and Framing Analysis, enabling the exploration of different perspectives presented in the media.

APPENDIX

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