Framework for NLP-Driven Insights into Patient-Centered Communication in Oncology

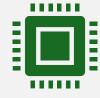
Grace Donovan

MSDE/MSDS Practicum 1 Regis University, Denver CO June 26, 2025

Overview



Introduction: Background, Objective, and Key Questions



Framework overview: Data Retrieval, Data Processing, and Data Analysis



Conclusion: Insights, Future Improvements



Background

1

Patient-Centered Care

Patient-focused communication is vital - driving treatment adherence, boosting satisfaction, and improving health outcomes (Sharkiya (2023), Krist et al. (2017), Becker, C., et al. (2021)).

2

Research Challenges

Synthesizing evolving communication practices and associated attitudes from the rapidly expanding medical literature can be difficult, complicating efforts to uphold high-quality patient care (Links, M., et al. (2020)).

"Effective communication is a cornerstone of quality healthcare. Communication helps providers bond with patients, forming therapeutic relationships that benefit patient-centered outcomes."
- Samer H. Sharkiya (2023)

3 Automated Insights

Automation of literature harvesting and analysis, can provide clinicians, researchers, and policymakers with real-time, data-driven decision support (van Dinter, R., et al. (2021)).



Objective

Develop framework using opensource tools to automate the synthesis and analysis of literature in an evolving research landscape



Topical Focus

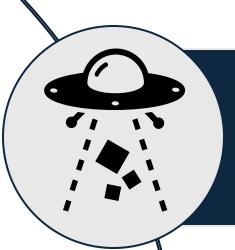
Exploring patient-centered communication themes in peer-reviewed oncology literature



Key Questions

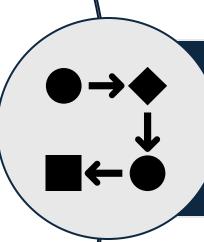
- 1. Which patient-centered communication topics emerge most prevalent in oncology research?
- 2. How does author sentiment vary across these topics?

Framework Overview



1. Data Retrieval

- Related publication identification
- Publication full-text data scraping



2. Data Processing

- Cleaning, standardizing, sub-setting, and transforming scraped HTML data to JSON dictionary
- Preprocessing cleaned text data for analysis (stop word removal, POS tagging, lemmatization, etc.)



3. Data Analysis

- Exploratory data analysis using word frequency distributions and n-grams
- Aggregating topic modeling for theme identification and sentiment analysis for sentiment classification of identified themes

Related Publication Identification and Initial Retrieval using xDD API

Component 1 – Data Retrieval

What is xDD?

• A cyberinfrastructure that compiles data on published literature and provides users with the ability to perform full text searches of published literature through the xDD API (Peters et al. 2025).

Workflow:

- Keyword Query: publications containing both "patient-centered communication" and "cancer" keywords
- **Filter:** articles with ≥ 2 total keyword hits, published between 2020 and 2025.
- Extracted Fields: Publication name, publisher name, title, date, DOI
 (Digital Object Identifier), author names, text snippets ("highlights")
 where keywords appeared, and count of keyword occurrences per
 document
- Output: Consolidated 130 publication metadata entries into a single JSON file for downstream analysis or visualization.

```
1_data_retrieval > data > {} xdd_pubs.json > {} 128 > # hits
               "pubname": "Supportive Care in Cancer",
               "publisher": "Springer",
               "title": "Oncology patients\u2019 communication experiences during COV
               "publication_date": "2022 06",
               "doi": "10.1007/s00520-022-06897-8",
               "authors": "Street, Richard L; Treiman, Katherine; Kranzler, Elissa C.
               "highlight": [
                    "The <em class=\"hl\">Patient</em>-<em class=\"hl\">Centered</em>
                    "<em class=\"hl\">Patient</em>\u2011<em class=\"hl\">centered</em>
 12
                    "<em class=\"hl\">Patient</em>-<em class=\"hl\">centered</em> <em</pre>
                    "<em class=\"hl\">Patient</em>-<em class=\"hl\">centered</em> <em</pre>
 14
                    "Psychometric evaluation and design of <em class=\"hl\">patient</e
               "hits": 5
 17
               "pubname": "Supportive Care in Cancer",
               "publisher": "Springer",
 21
               "title": "Hodgkin lymphoma survivor perspectives on their engagement i
 22
               "publication_date": "2022 02",
 23
               "doi": "10.1007/s00520-021-06538-6",
 24
               "authors": "Murphy-Banks, Rachel; Kumar, Anita J.; Lin, Mingqian; Savi
 25
               "highlight": [
                    "Implications for <em class=\"hl\">cancer</em> survivors\u2002 <em
 27
                    "Keywords\u2002 Late effects \u00b7 Hodgkin lymphoma \u00b7 <em cl
 28
                    "<em class=\"hl\">Patient</em>-<em class=\"hl\">centered</em> <em</pre>
 29
                    "This contributes to our position that more attention on <em class
                   "Epstein RM, Street RL, Jr (2007) <em class=\"hl\">Patient</em>-<e
 31
 32
               "hits": 5
 33
 34
               "pubname": "Journal of Perinatology",
 36
               "publisher": "Springer",
 37
               "title": "Communication between neonatologists and parents when progno
               "publication_date": "2020 09",
               "doi": "10.1038/s41372-020-0673-6",
               "authors": "Drach, Laura L.; Hansen, Debra A.; King, Tracy M.; Sibinga
 41
               "highlight": [
 42
                    "Study design Guided by the National <em class=\"hl\">Cancer</em>
                    "<em class=\"hl\">centered</em> <em class=\"hl\">communication</em</pre>
 43
                    "The National <em class=\"hl\">Cancer</em> Institute (NCI) <em cla
                    "The <em class=\"hl\">Patient</em> <em class=\"hl\">Centered</em>
                    "<em class=\"hl\">Patient</em>-<em class=\"hl\">centered</em> <em
 47
               "hits": 5
```

Full-text Data Scraping of Related Publication Webpages

Component 1 – Data Retrieval

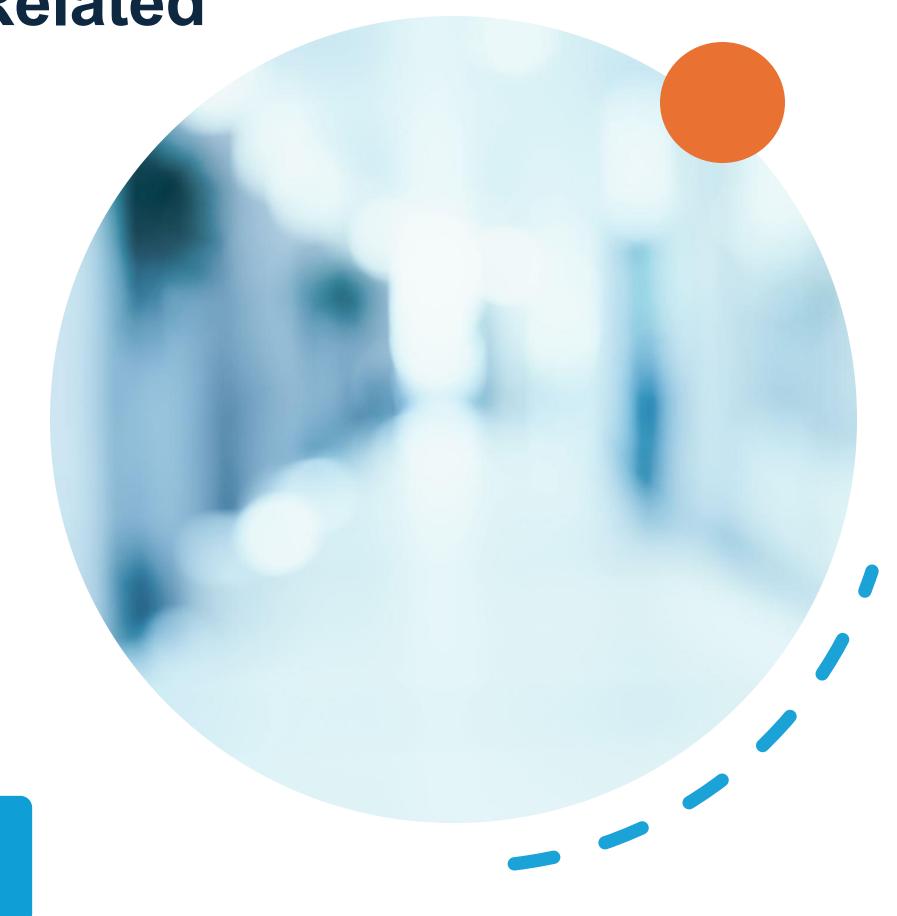
Constructed publication DOI URLs from DOI identifiers collected from xDD API

Used Selenium WebDriver and BeautifulSoup for scraping full-text from publication webpages

Specific sections targeted for scraping (abstract, methods, results, conclusion, etc.)

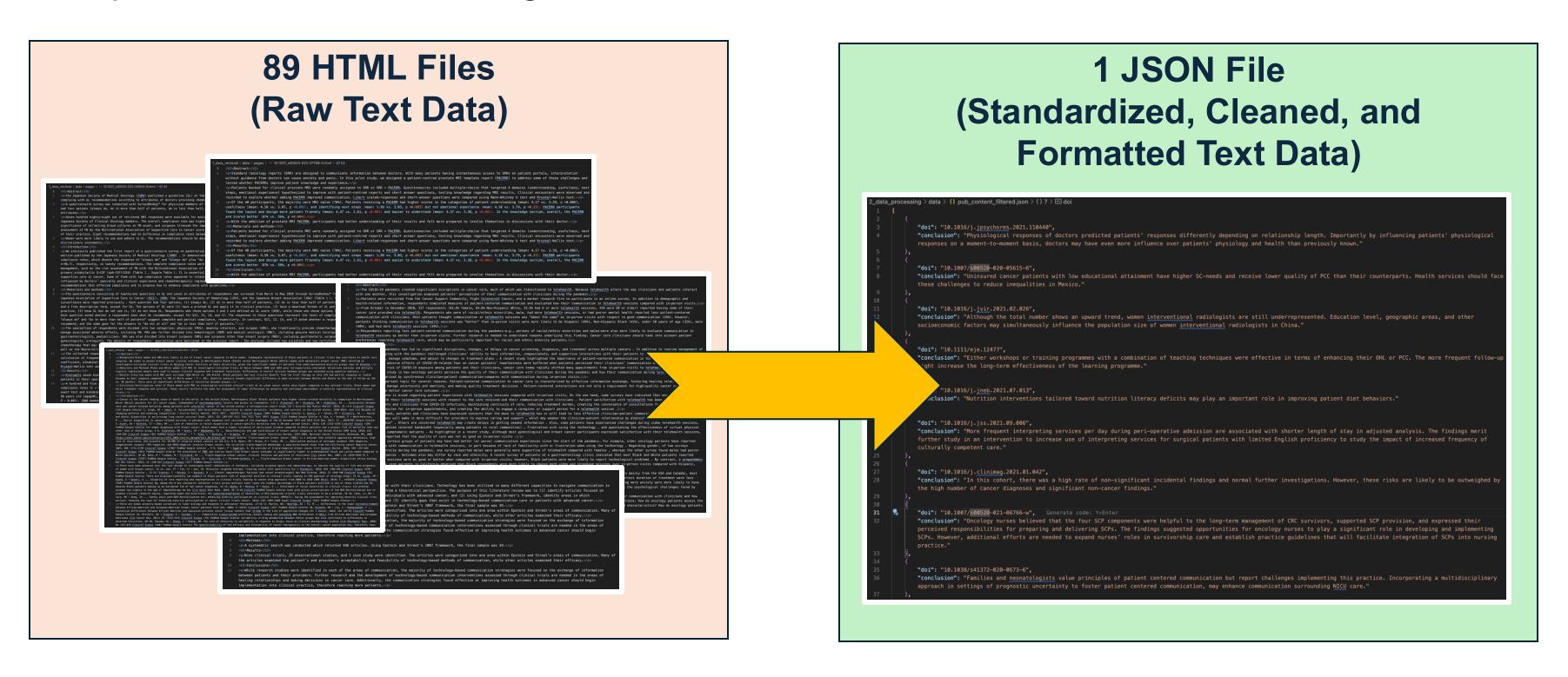
Filters included for handling inconsistencies and scraping issues

89 publications successfully scraped and saved as HTML files



HTML Content Extraction and Cleanup

Component 2 – Data Processing



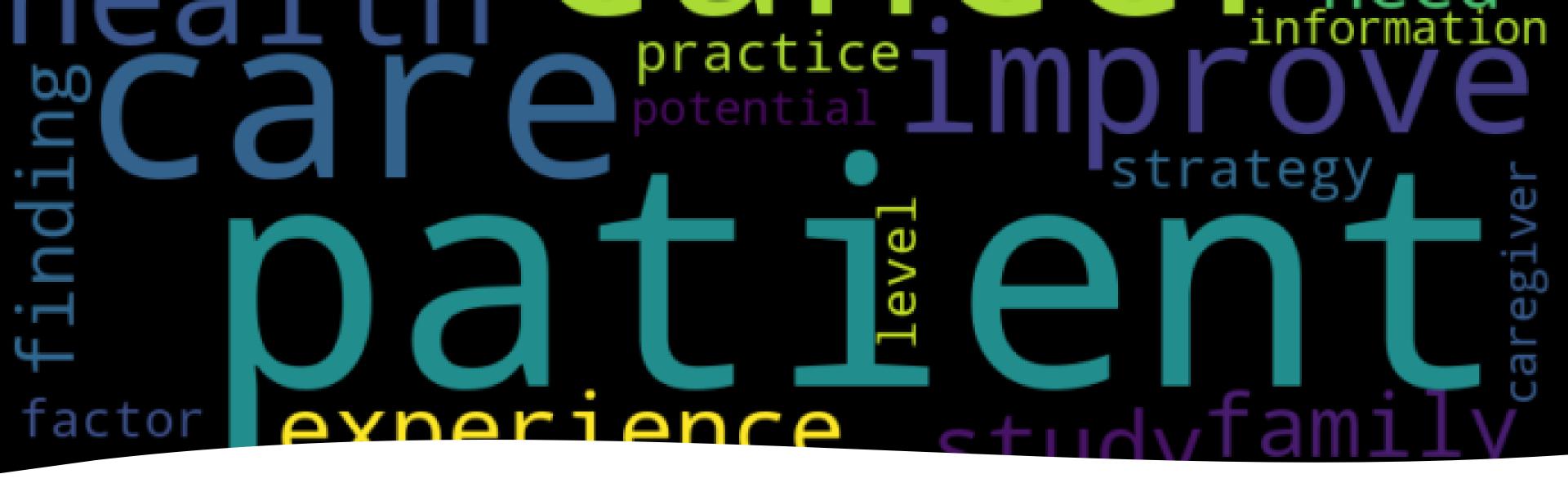
Preprocessing Conclusion Data for Analysis

Component 2 – Data Processing

Process:

- **Tokenization**: Sentence → word level
- Lowercasing: Normalize to lowercase
- Stop word removal: stop words + domain-specific terms
- Length filtering: Remove words <3 characters
- POS filtering: Remove adverbs (less contextually relevant)
- Lemmatization: Reduce words to root forms





Exploratory Data Analysis

Component 3 – Data Analysis

Word Frequency Analysis

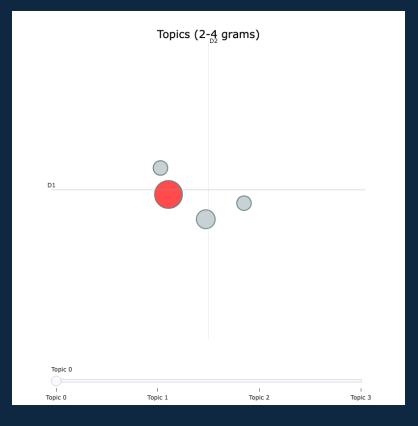
- Top terms: patient (54), care (25), cancer (24), communication (22)
- "Patient" occurrence confirmed research focus alignment

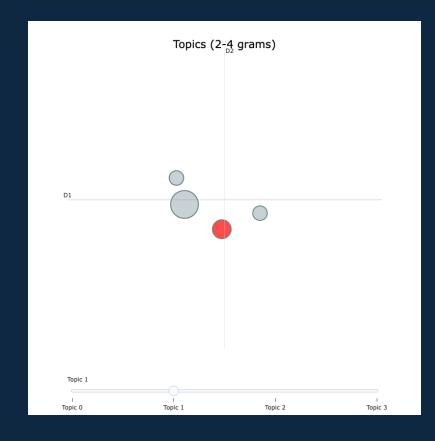
N-gram Pattern Dcanceriscovery

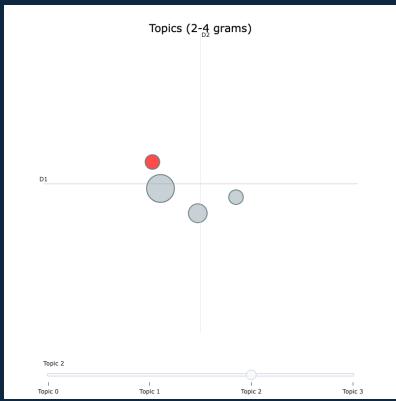
- Bigrams: "cancer patient" (5), "cancer care" (4), "family caregiver" (4), "lung cancer" (4), "health information" (3)
- Trigrams: "health information platform" (2), "cancer communication guideline" (2), "patient support person" (2), "positive health outcome" (2), "experience racial discrimination" (2)
- "Patient" occurrence confirmed research focus alignment

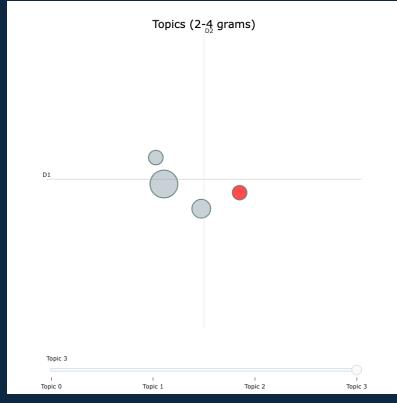
Topic Modeling Implementation

Component 3- Data Analysis









BERTopic Clustering:

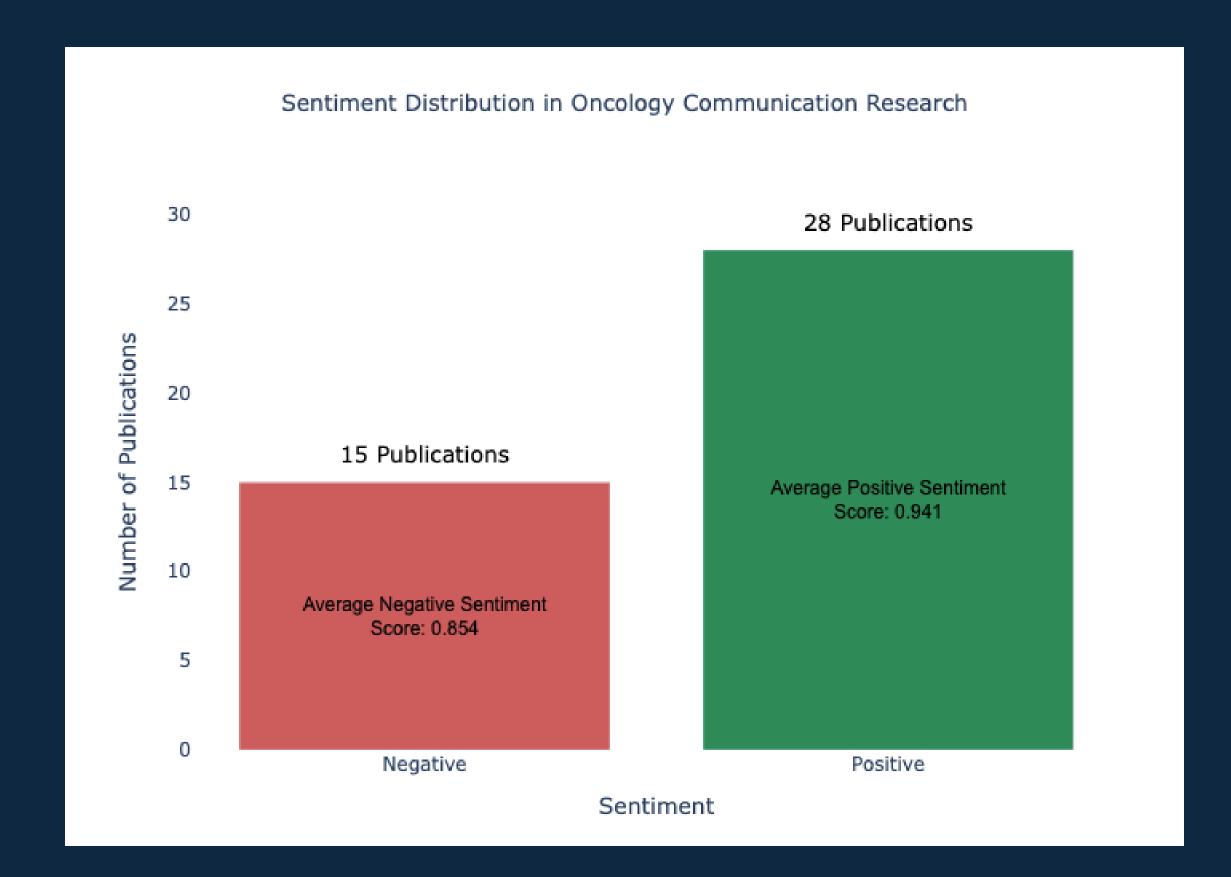
- Multi-word theme extraction using n-grams (2-4 word phrases)
- UMAP dimensional reduction for semantic clustering
- HDBSCAN density-based topic formation
- Minimum cluster size: 3 documents (coherent topics)

Sentiment Analysis

Component 3 - Data Analysis

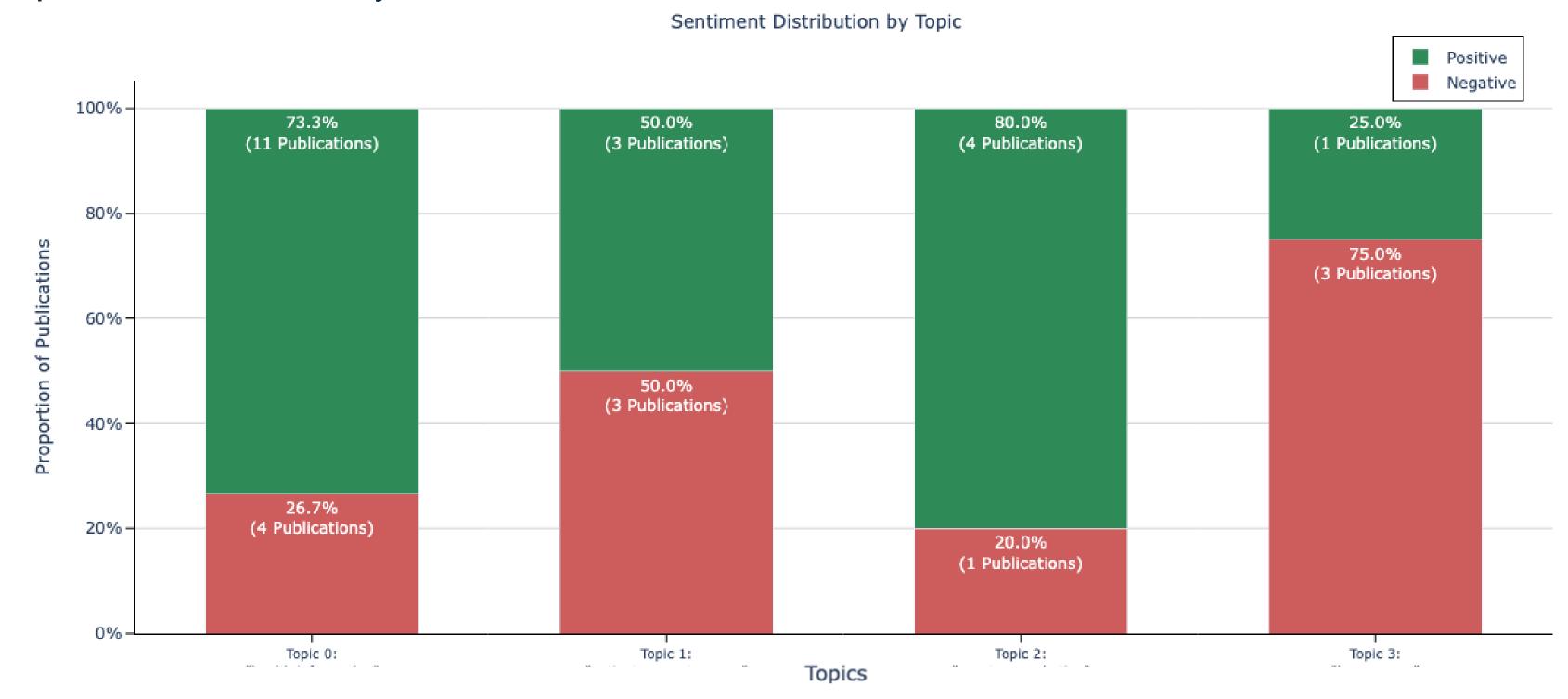
- DistilBERT model

 (distilbert-base-uncased-finetuned-sst-2-english)
 applied to publication conclusions
- Predominantly positive sentiment identified, revealing research community attitudes toward interventions



Aggregation of Topic-level Sentiment Distribution Analysis

Component 3 - Data Analysis



Aggregation of Topic-level Sentiment Distribution Analysis

Component 3 - Data Analysis

Topic 0 Top Keywords:

- "health information"
- "secure message"
- "improve patient"
- "oncology nurse"

Topic 1 Top Keywords:

- "patient support person"
- "cancer survivor"
- "support person"
- "experience racial discrimination"

Topic 2 Top Keywords:

- "onset exacerbation"
- "therapeutic burden"
- "patient crpc"
- "drug product"

Topic 3 Top Keywords:

- "lung cancer"
- "acute traumatic"
- "radiotherapy advance"
- "mitigate risk adverse"



Project Insights and Lessons Learned

- xDD streamlines discovery of related studies and automates retrieval of publication metadata
- Capturing full-text articles remains challenging, due to publisher paywalls and inconsistent website structures
- Multi-word extraction more effectively captured relevant terms
- A combined topic modeling and sentiment analysis approach revealed both optimistic and cautionary tones in key themes

Future Improvements

- Performance & Monitoring: Refined and async scraping
- Model & Research: Recalibrate models, add temporal, cancer-type, and outcome analyses, and analyze more publication content (other than conclusions).
- **Deployment**: Migrate to a database backend



Conclusion



Approach

- Automated pipeline for literature mining and interpretable outputs
- Combined NLP methods (BERTopic, DistilBERT)



Key Findings

- Identified dominant themes in patient-centered cancer communication
- Detected mixed sentiment toward communication-focused interventions



Research Impact

- Offers an evidence-based framework for identifying oncology communication strategies
- Has the potential to reveal best practices that can be consolidated into a unified approach



Technical Contribution

- Architecture tailored to midsize scientific literature corpora
- Modular components could be reused across domains

References

- Becker, C., et al. Interventions to Improve Communication at Hospital Discharge and Rates of Readmission: A Systematic Review and Meta-analysis. JAMA network open vol. 4,8 e2119346. (2021), https://doi.org/10.1001/jamanetworkopen.2021.19346.
- Krist, A.H., et al. "Engaging Patients in Decision-Making and Behavior Change to Promote Prevention." Studies in health technology and informatics vol. 240 (2017), https://pmc.ncbi.nlm.nih.gov/articles/PMC6996004/.
- Links, M.J., et al. Finding common ground: meta-synthesis of communication frameworks found in patient communication, supervision and simulation literature. BMC Med Educ 20, 45 (2020). https://doi.org/10.1186/s12909-019-1922-2.
- Peters, S.E., et al. xDD: About. (2025), https://geodeepdive.org/about.html.
- Sharkiya, S.H., Quality communication can improve patient-centred health outcomes among older patients: a rapid review. BMC Health Serv Res 23, 886 (2023). https://doi.org/10.1186/s12913-023-09869-8.
- van Dinter, R., et al. A decision support system for automating document retrieval and citation screening, Expert Systems with Applications, (2021), https://doi.org/10.1016/j.eswa.2021.115261.
- All images and icons used in the presentation were sourced from PowerPoint.

Thank you!