

# Automatic Article Generation via Multi-Document Summarization

Connor Lennox  
connor.lennox@unh.edu  
Department of Computer Science, UNH

Advisor: Laura Dietz


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## Information Retrieval

- ▶ When a user (you) enters a query (web search), what documents (pages) should we display to provide information?

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Horseshoe Crab × 

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About 253,000 results (0.40 seconds)

[https://en.wikipedia.org > wiki > Horseshoe\\_crab](https://en.wikipedia.org/wiki/Horseshoe_crab) ⋮

## Horseshoe crab - Wikipedia


**Horseshoe crabs** are marine and brackish water arthropods of the family Limulidae and the only living members of the order Xiphosura.

Family: Limulidae; Leach, 1819 Kingdom: [Animalia](#)  
Phylum: [Arthropoda](#) Order: [Xiphosura](#)  
[Atlantic horseshoe crab](#) · [Limulus amebyocyte lysate](#) · [Mangrove horseshoe crab](#)

[https://www.nwf.org > Wildlife-Guide > Invertebrates](https://www.nwf.org/Wildlife-Guide/Invertebrates) ⋮

## Horseshoe Crab | National Wildlife Federation

**Horseshoe crabs** have been around for more than 300 million years, making them even older than dinosaurs. They look like prehistoric crabs, but are actually ...



[https://www.chesapeakebay.net > fieldguide > critter > hor...](https://www.chesapeakebay.net/fieldguide/critter/hor...) ⋮

## Horseshoe Crab | Chesapeake Bay Program

**Horseshoe crabs** grow to two feet in length. They have a hard, rounded, brownish-green exoskeleton, a spike-like tail and five pairs of jointed legs. Their ...

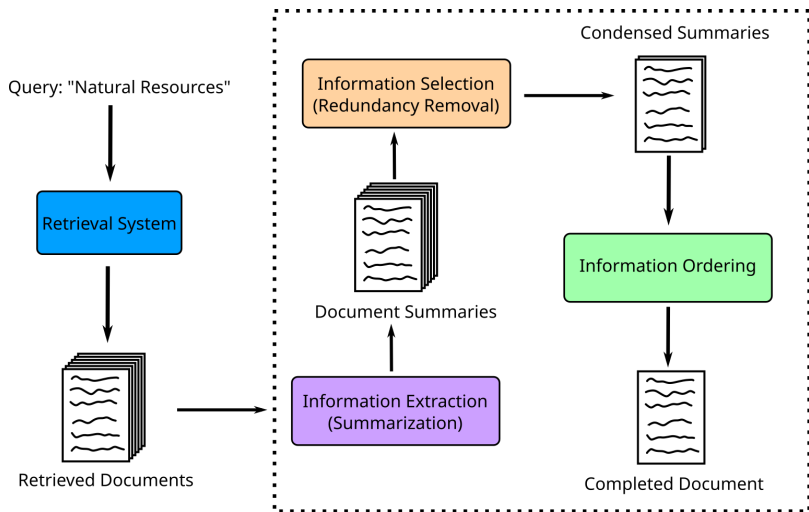
# Task

- ▶ Users don't want to read an entire ranking of documents to find the information they're looking for!
- ▶ Goal: Develop a system that can effectively merge all documents of a ranking into a single, informative article which can be presented to the user
- ▶ **Given query  $q$  and ranking of documents  $d_1, d_2, \dots, d_n$ , generate an article  $a$  that is similar to a reference article.**

# Challenges

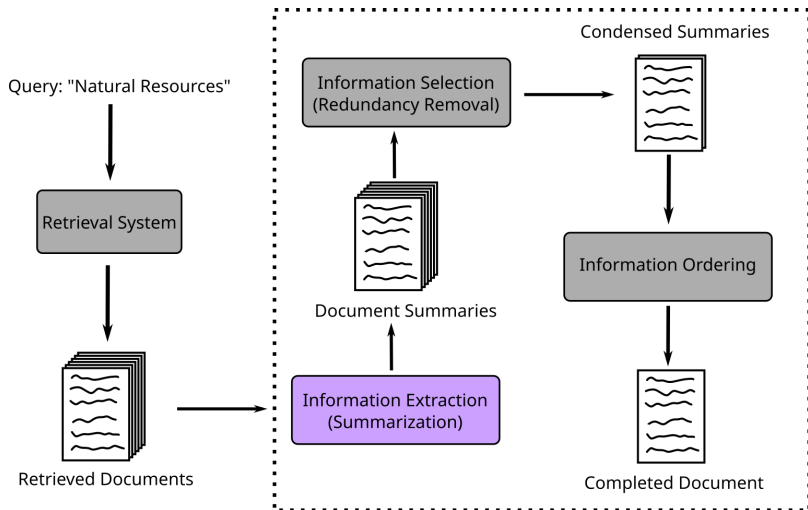
- ▶ **Text Quantity:** Document rankings contain a large amount of text, too much to process simultaneously
- ▶ **Topical Breadth:** For broad queries, many subtopics will be addressed in documents
- ▶ **Information Ordering:** It is important that information is presented to the user in a logical, “coherent” order

# Generation Pipeline

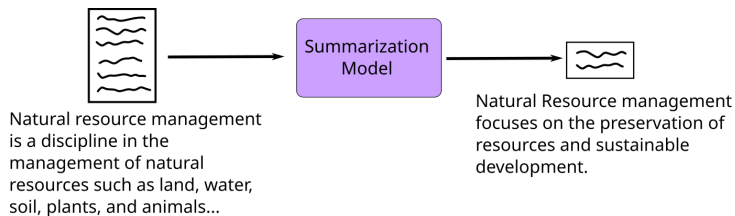


**Figure:** The full article generation pipeline. Elements within the dashed box represent those we will address in this work.

# Information Extraction



# Information Extraction



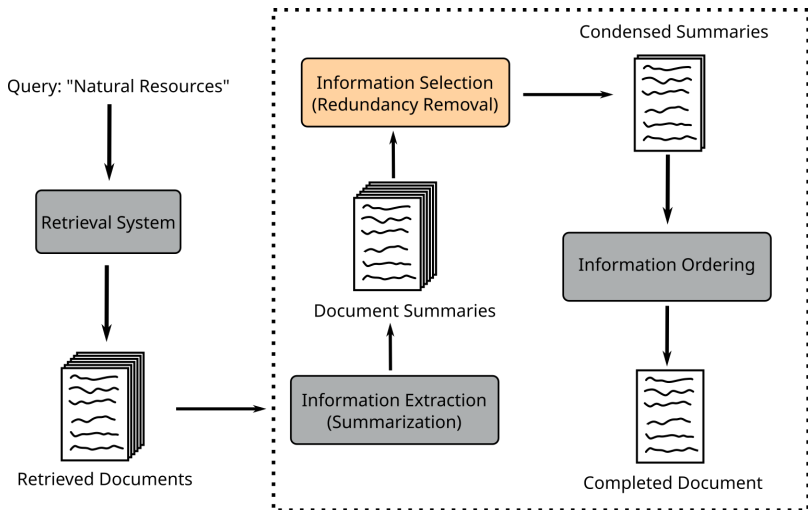
**Figure:** From an input document, information can be extracted through the use of a single-document summarization model. Here, some facts about the management of Natural Resources are found in source text to produce a shorter summary.



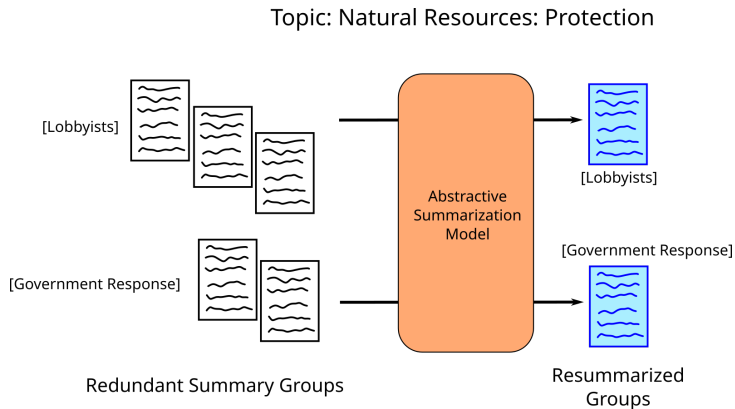
# Information Extraction

- ▶ Each input document will have some information we want to extract and add to our final article
- ▶ Modern summarization models achieve great performance on *single-document summarization*, leverage this to extract core information of each input document
- ▶ By applying single-document summarization to all input documents, we are left with a collection of facts that we want to include in our final article

# Information Selection



# Information Selection



**Figure:** By grouping together redundant information and resummarizing, a more concise output article can be constructed

## Information Selection

- ▶ By clustering information, we can identify redundant facts and combine them into a single summary
- ▶ Eliminating redundancy allows the final output to be more concise, without risk of completely removing critical information from the text
- ▶ Redundant groups are merged by using the same summarization model that performs information extraction

## Discussion

- ▶ Our method of article generation avoids common pitfalls by splitting the input into smaller subproblems
- ▶ Components are modular and interchangeable: the system is agnostic to the exact summarization model used, etc.
- ▶ A side effect of our method is the ability to derive provenance of every statement in the output document

# Evaluation

- ▶ Comparisons are made between a heuristic baseline method as well as two methods from recent literature
  - ▶ **Baseline** No summarization, input is output
  - ▶ **Hierarchical Transformer** (Liu et al., 2019)
  - ▶ **LoBART** (Pasunuru et al., 2021)

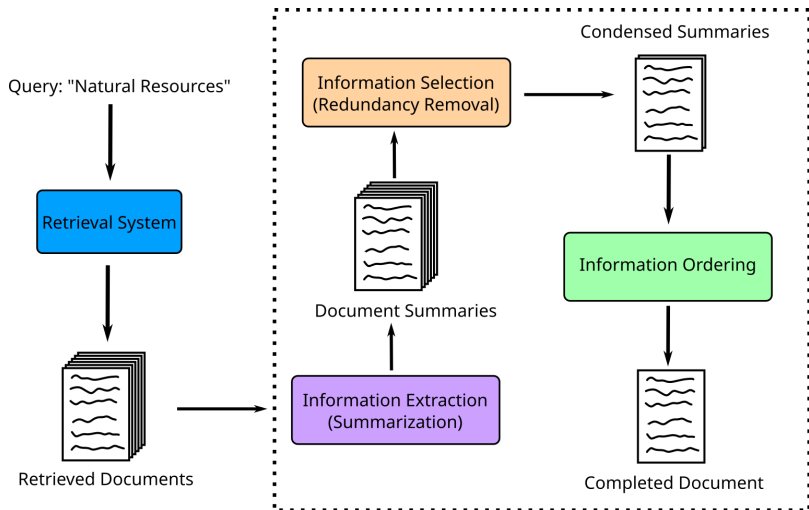
# Results

Model	ROUGE-1	ROUGE-2	Manual Evaluation
Heuristic Baseline	$0.165 \pm 0.004$	$0.027 \pm 0.002$	$1.44 \pm 0.17$
Hierarchical Transformer	$0.074 \pm 0.005$	$0.013 \pm 0.001$	-
LoBART	$0.211 \pm 0.005$	$0.052 \pm 0.002$	$1.80 \pm 0.14$
Multistage (Ours)	$0.172 \pm 0.003$	$0.028 \pm 0.001$	$1.54 \pm 0.16$

**Table:** Results of evaluation with both automatic and manual metrics. Manual scores are on a range of 0-3. For all metrics, higher is better.

- ▶ **ROUGE:** Evaluates *linguistic* overlap
- ▶ **BERTScore:** Evaluates *semantic* overlap
- ▶ **Manual Evaluation:** Annotators asked to evaluate presence of “important information”

# Conclusion





## Conclusion

- ▶ We have proposed a system for summarizing a set of documents obtained via an Information Retrieval system.
- ▶ Our method, Multi-Stage Cluster Summarization, avoids possible challenges by breaking down the summarization problem into subtasks.
- ▶ The system is capable of producing informative yet concise documents that encompass the full breadth of information about a topic.