# BIASsist = Bias + Assist

**Empowering News Readers via Bias Identification, Explanation, and Neutralization** 



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MinJu Han

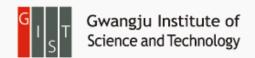


Junryeol Jeon



Jin-Hyuk Hong



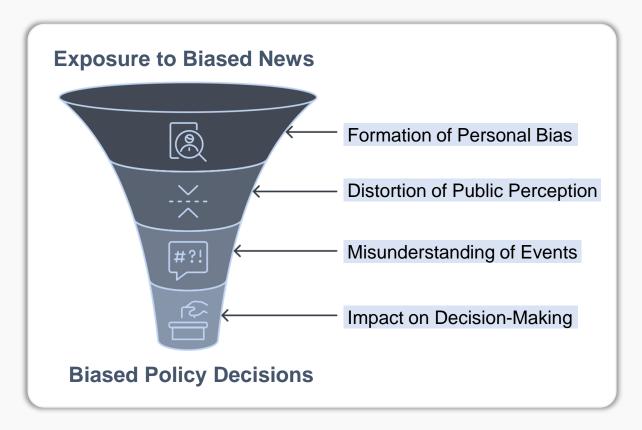




### **Introduction & Motivation**

Biased news articles can distort readers' perceptions

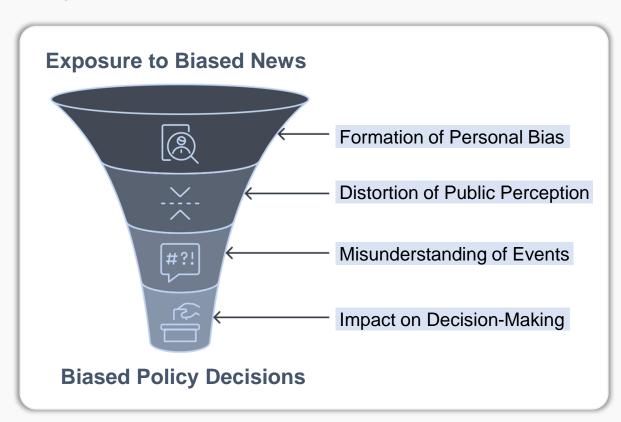
Negative Consequences of Exposure to Biased News



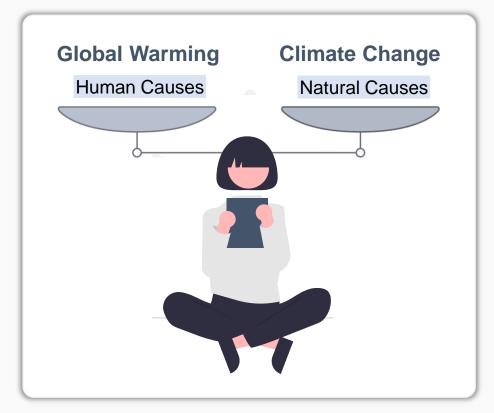
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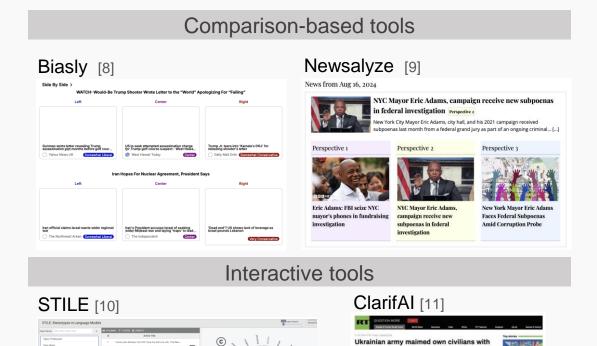


The Impact of Terminology on Perception



## **Related Works**

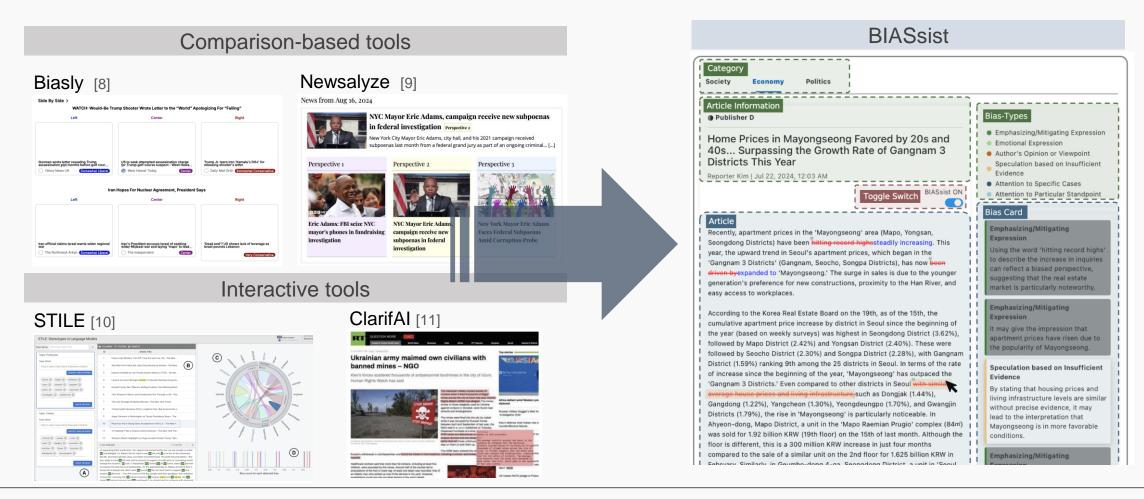
Many tools assist with article comparison or search but offer limited explanation of various bias types



banned mines - NGO

### **Related Works**

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## **Contributions**

The key contributions of this work are:

#### (1)BIASsist Development

: Designed an LLM-based tool to enhance bias identification, explanation, and neutralization across six bias types.

#### (2) Empirical Insights

: Conducted a **mixed-method study** to assess user perception changes and validate the tool's effectiveness.

#### (3) Implications & Future Research

: Highlighted the **tool's role** in fostering critical reading and discussed **potential concerns**.

**Six Types of Bias** & Three Assistive Components **Word Choice EMX** #?! / Expression Emphasizing/Mitigating Expression EmoX **Emotional Expression** Authors' **AOV** Involvement Author's Opinion or Viewpoint SpecIE Speculation based on Insufficient Evidence Selective **ASC** Attention to Attention to Specific Cases Sources **APS** Attention to Particular Standpoint

### Example of EMX

Biased This incident is a major disaster caused by the government's severe incompetence.

Neutral This incident is evaluated as a challenge in the government's response.

#### Example of AOV

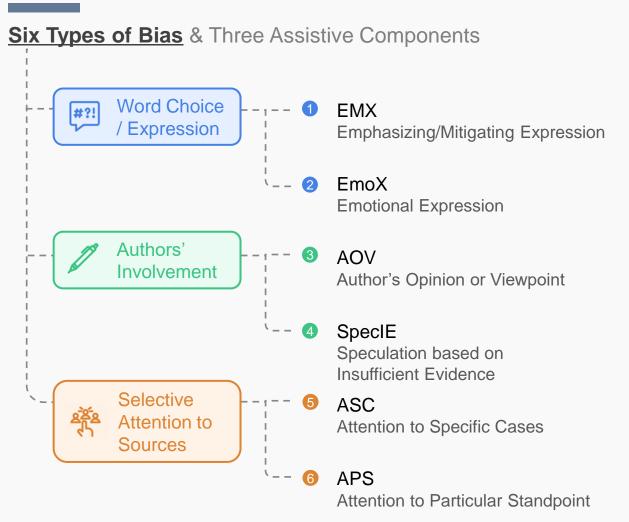
Biased Thib bill will certainly make people's lives more difficult.

Neutral Analyses by experts have rasied concerns about the impact of this bill on peoples's lives.

#### Example of ASC

Biased The drunk driving of athlete A caused a social uproar. Similary, last year, athelet B and C also...

Neutral The drunk driving of athlete A caused a social uproar.



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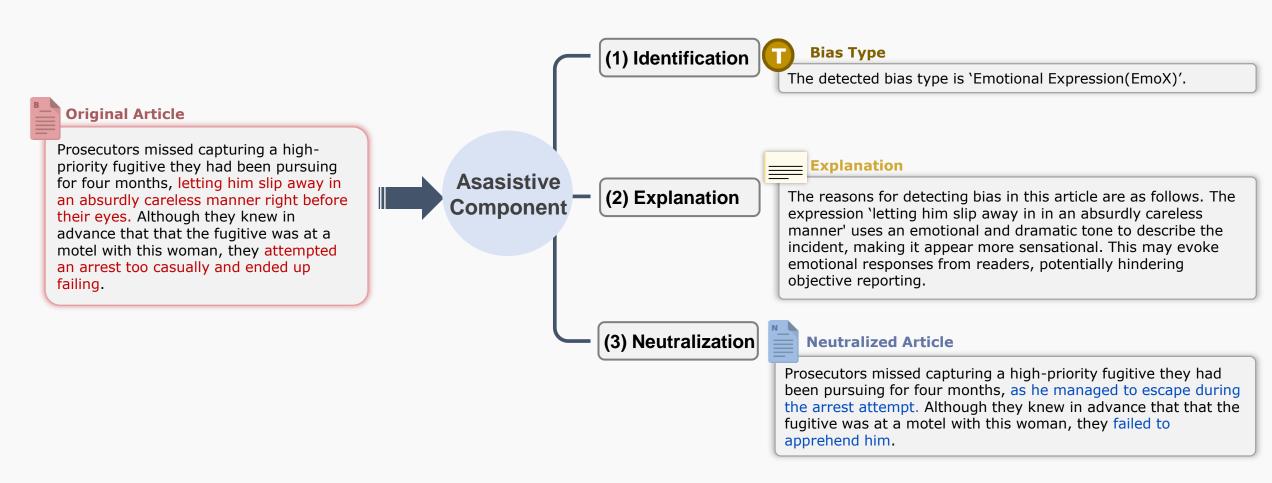
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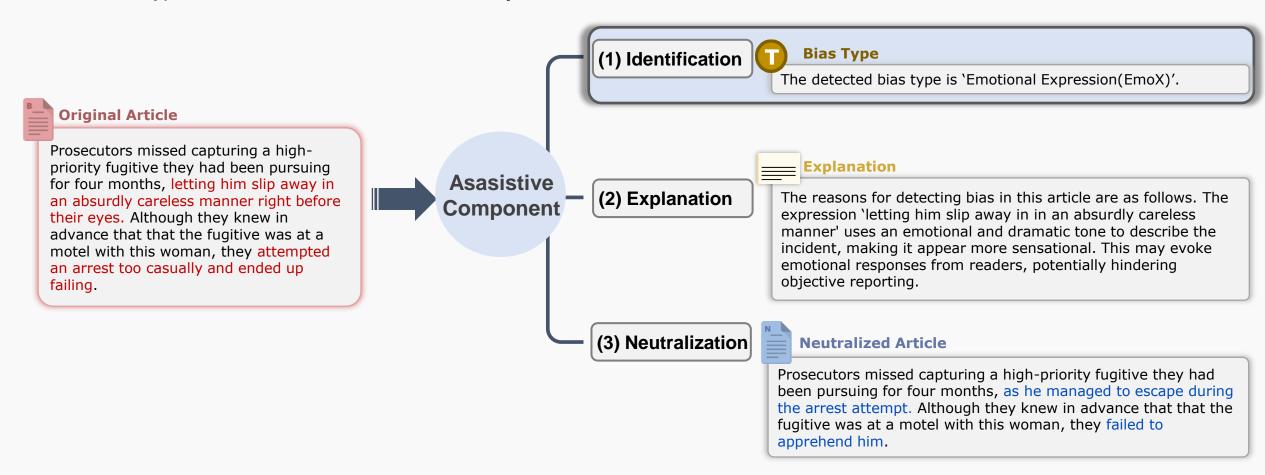
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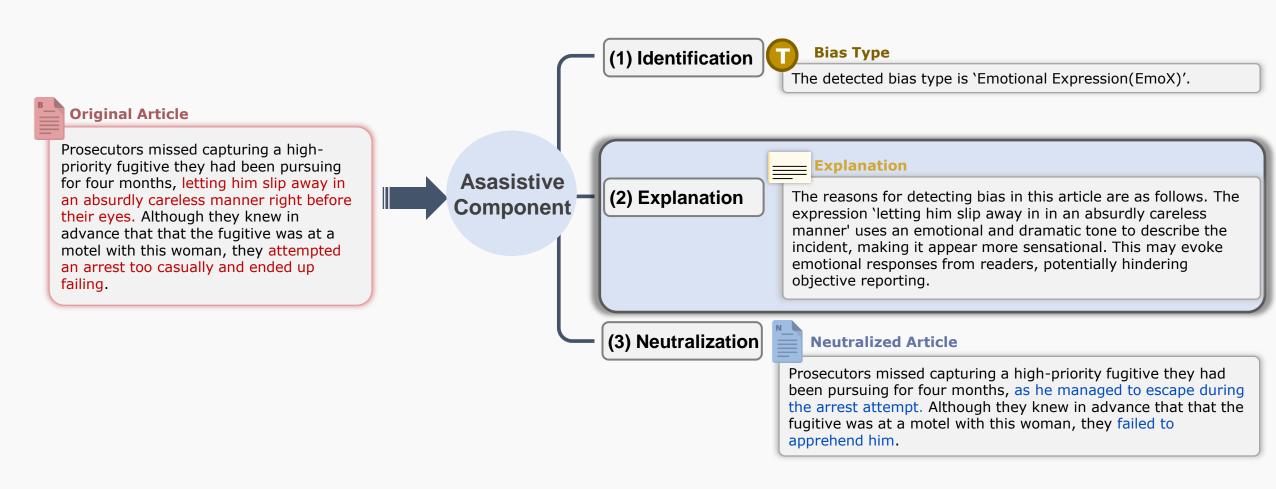
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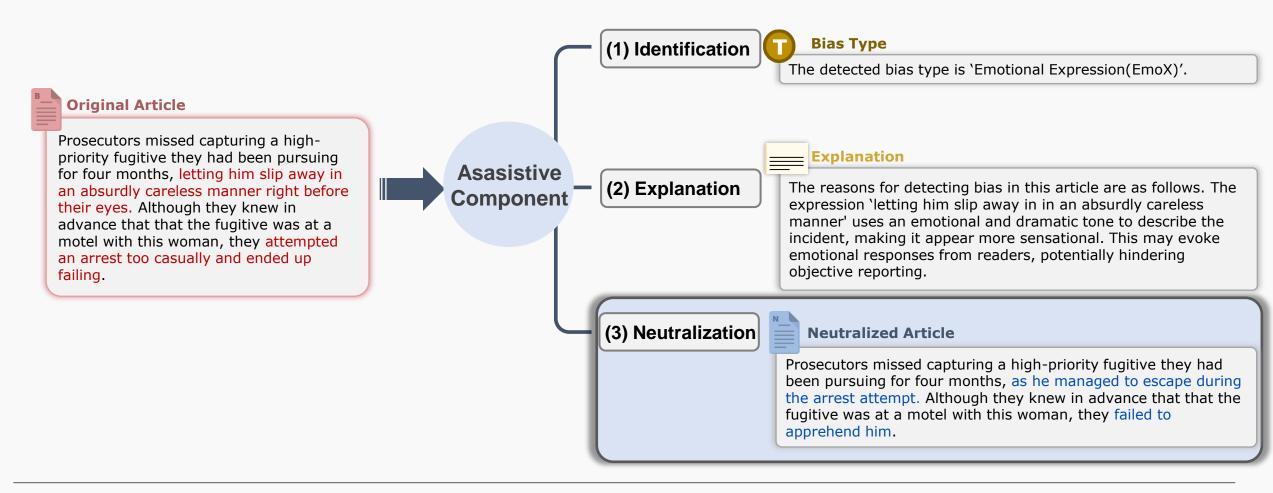
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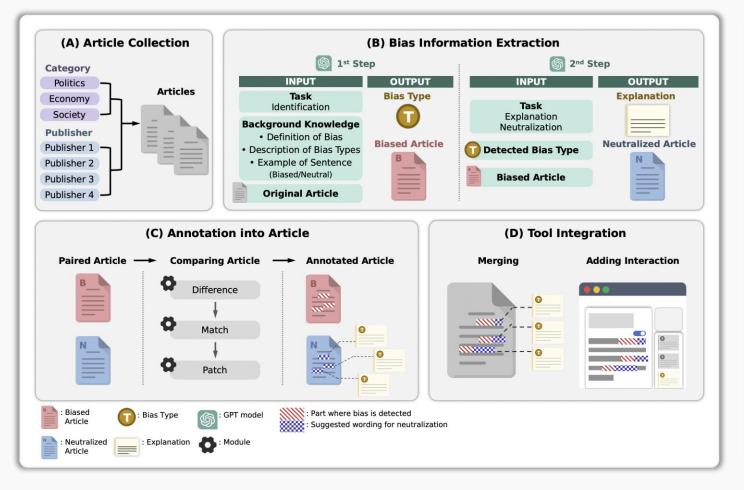
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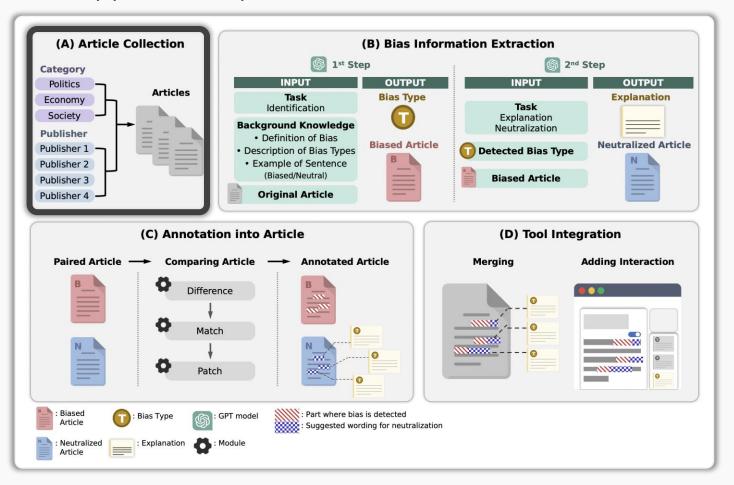




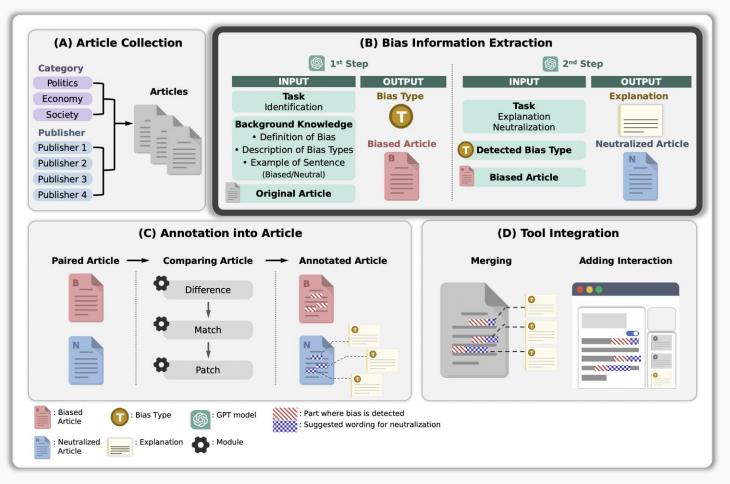




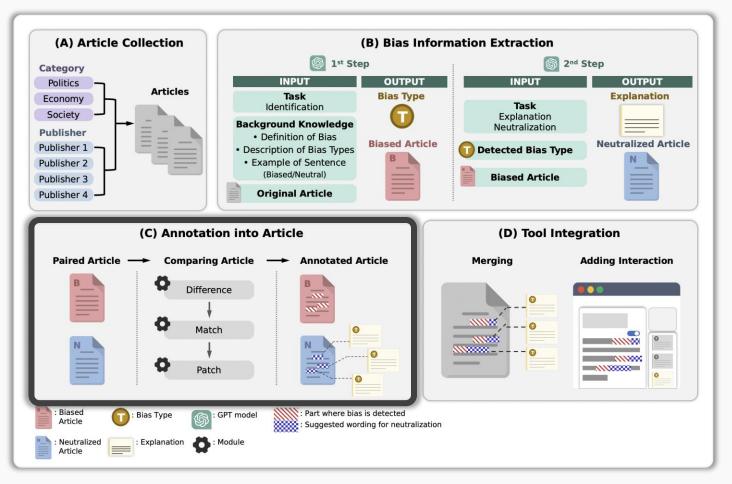




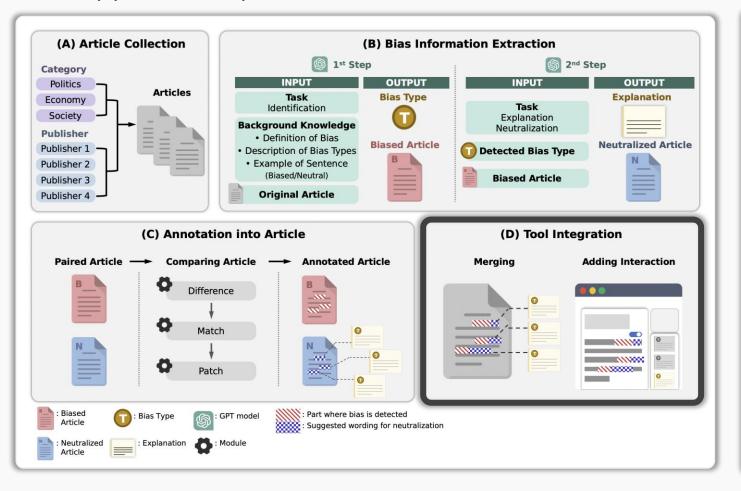










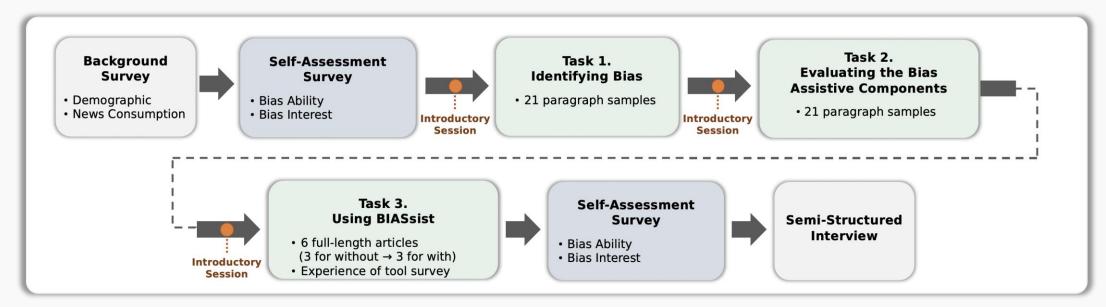




# **User Study**

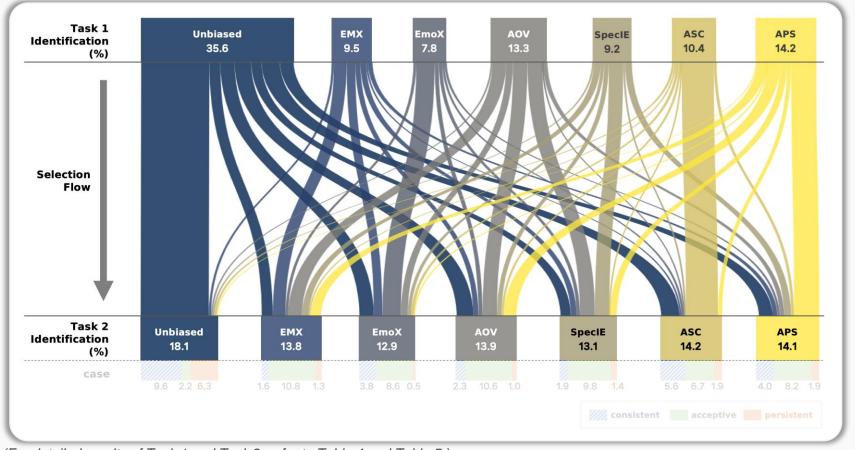
- Study Design
- Within-subjects design
- 36 participants recruited
  - age from 18 to 55 years (M=25.3, SD=8.2)
  - 21 females (58.3%) and 15 males (41.7%)

#### Procedure



# Result 1 : Selection Flow between Initial Choice to Assisted Choice Task 1 Task 2

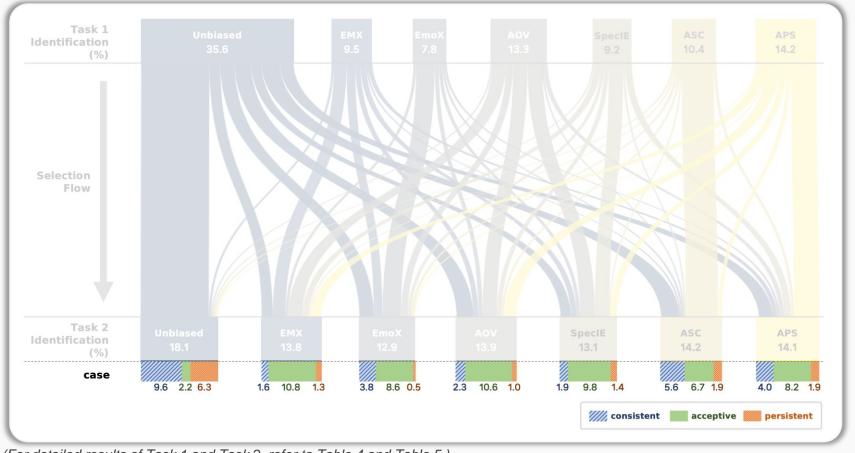
- Re-assessing the bias in articles initially judged as unbiased reduced the proportion considered unbiased from 35.6% to 18.1%.



(For detailed results of Task 1 and Task 2, refer to Table 4 and Table 5.)

# Result 1 : Selection Flow between Initial Choice to Assisted Choice Task 1 Task 2

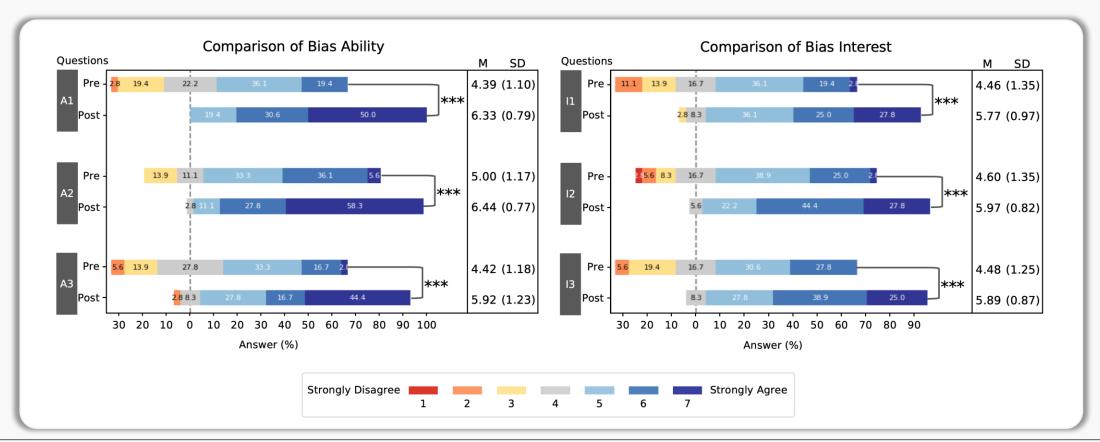
- Reflecting flexible opinion adjustments, three response categories emerged—consistent (28.8%), acceptive (56.9%), persistent (14.3%)



## Result 2: Bias Self-Assessment Survey

#### Comparison between pre and post experience

- Paired-t test showed significant differences (p < .001) across all items in both the Ability and Interest categories



# Result 3: Findings from the Interviews (1/2)

#### **Enhanced Bias Awareness & Perspective Broadening**

• Initially overlooked subtle biases – enhanced recognition through direct comparison with neutralized expressions.



P35 - "If you've already seen a biased article, it can be **hard to imagine a neutral expression** on your own. But seeing the same content expressed neutrally made it clearer to me."

#### **Balancing Content Understanding & Bias Detection**

• Simultaneously understanding content and detecting bias was challenging—BIASsist empowered both.



P4 - "When I didn't know much about the topic, I **focused too much on understanding** the content and couldn't think about bias. This tool has been really **helpful**."

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## Result 3: Findings from the Interviews (2/2)

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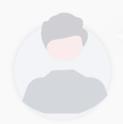
• Averting forced corrections and censorship—empowering user autonomy to critically assess



P6 - "I felt more **empowered** using the toggle because it reflected my intention to judge bias, rather than just passively reading the article."

### **Concerns & Improvements**

- Expert quote removal—raising information loss concerns.
- Suggestions include credibility scores and adjustable neutralization intensity.



P21 - "If it could tell me **numerically** how biased it is, I would trust the tool much more. An **absolute score** would also allow me to compare it with other articles"

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#### 1. Beyond Textual Bias

- Does not analyze headlines, images, or article layouts, which also influence perception.

#### 2. Real-World Application & Long-Term Study Needed

- Study conducted in a controlled environment with short-term user exposure.
- Requires longitudinal studies to assess long-term effects on media literacy

#### 3. Personalized Bias Neutralization

- Users have different tolerance levels for bias neutralization intensity.
- Future work: Allow users to adjust the level of rephrasing based on preferences.

- LLM-generated explanations and rewordings may introduce hallucinations.
- Retrieval-Augmented Generation (RAG) could help ground neutralization in factual sources.

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# Thank you for your attention



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