

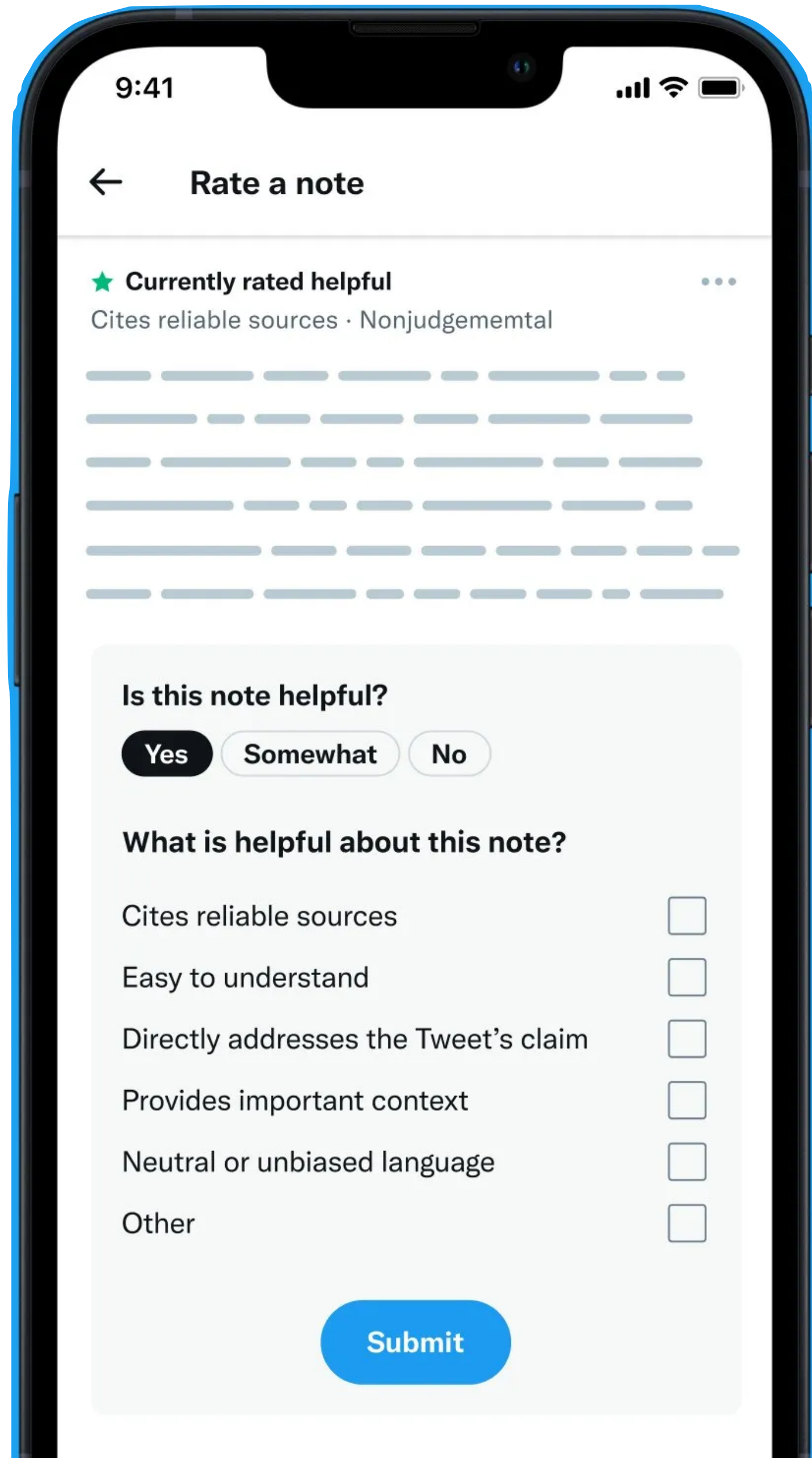
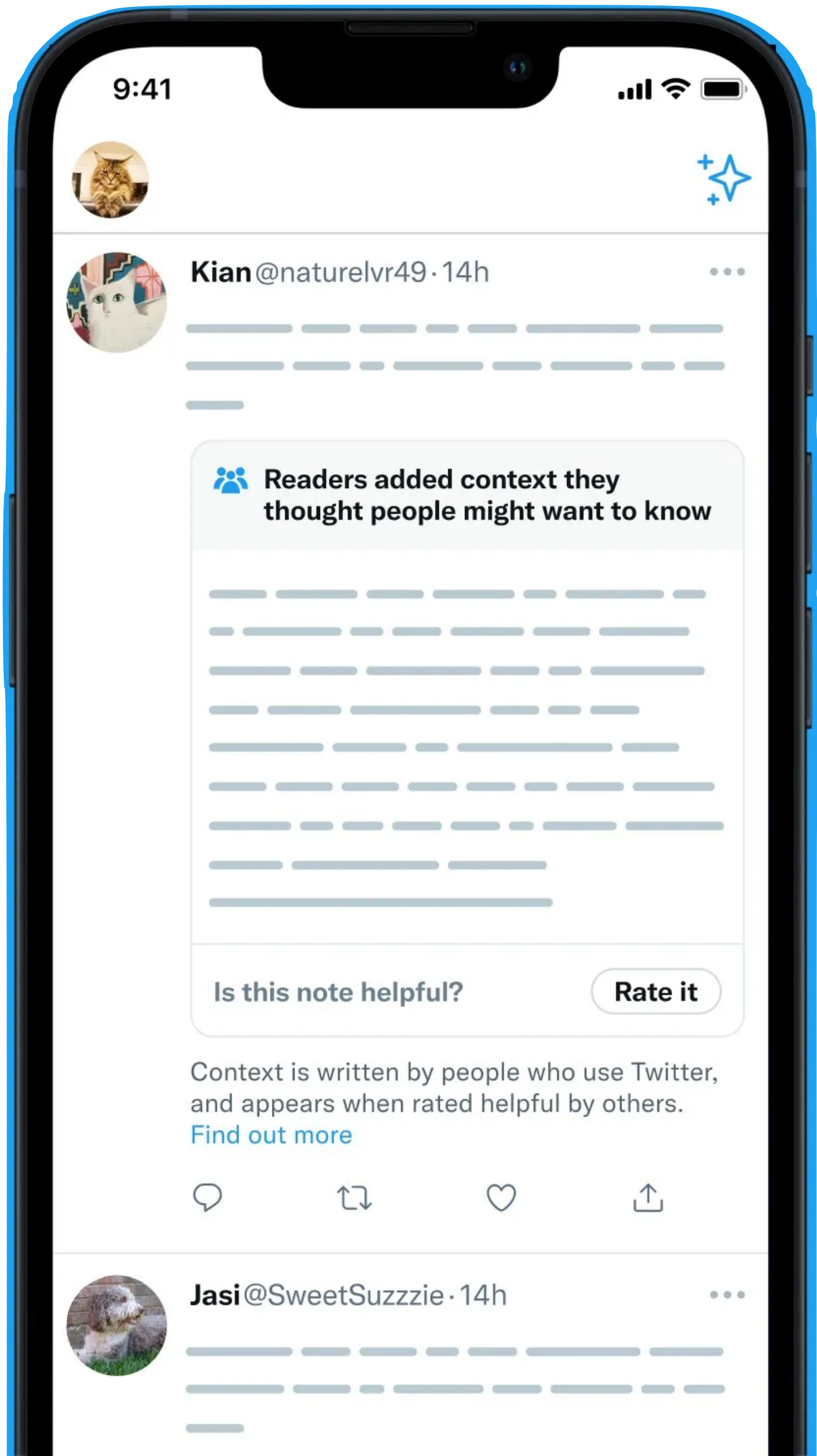
# The *Supernote*

Enhancing Crowd-sourced factchecking using  
LLM-driven consensus

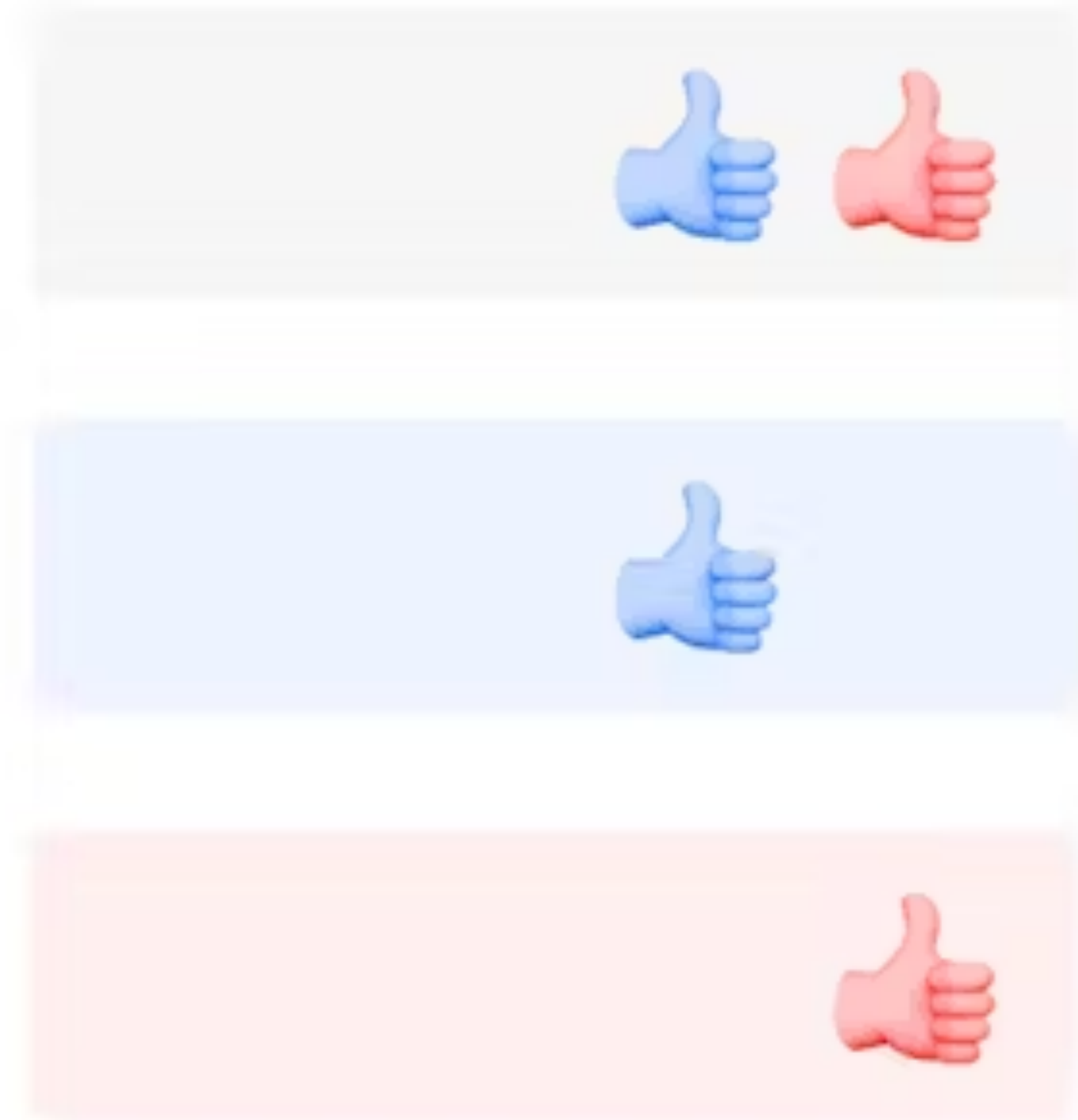
Soham De, Michiel Bakker, Martin Saveski

IC<sup>2</sup>S<sup>2</sup> 2024





# Community Notes



**Poynter Institute** on


*Why Twitter's Community Notes feature mostly fails to combat Misinformation*

**“It requires a cross-ideological agreement on truth, and ...  
achieving that consensus is almost impossible”**

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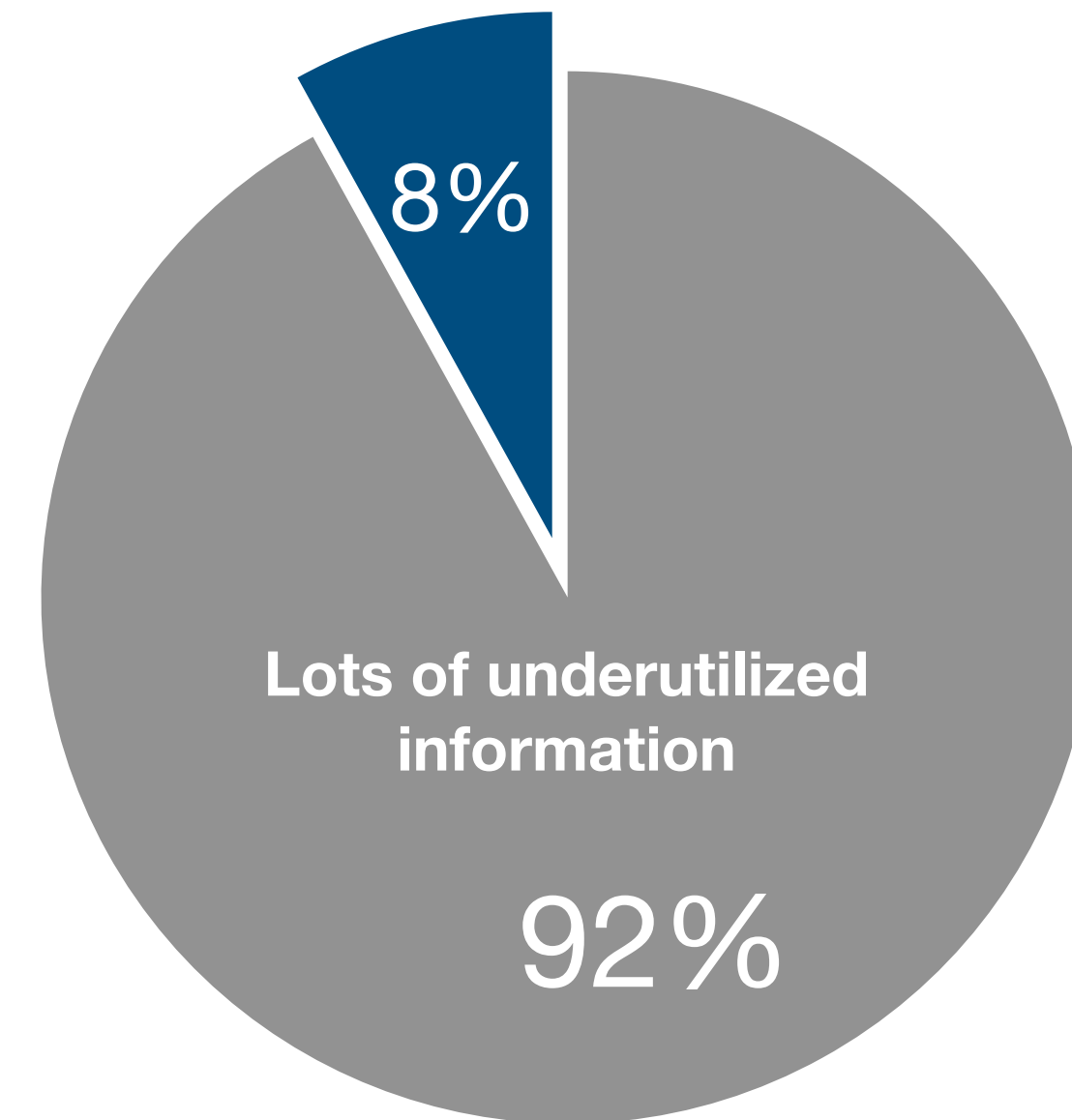


Notes aren't rated helpful fast enough to reduce engagement with misinformation in the early stages of diffusion

**(Chuai et al., 2023)**

**“It requires a cross-ideological agreement on truth, and ... achieving that consensus is almost impossible”**

Notes aren't rated helpful fast enough to reduce engagement with misinformation in the early stages of diffusion  
**(Chuai et al., 2023)**



as a consequence,

**More than 90% of notes written are not shown on X.**

“It requires a cross-ideological agreement on truth, and ... achieving that **consensus** is almost impossible”

Can you rephrase and summarize text well?



I'm so good at it!

LLMs can construct very good consensus statements to find agreement  
**(Bakker et al, 2022)**

# Goals

## 1. Improve fact-checking note-quality

- Use information in existing notes  
(useful information may be spread across several notes)
- Ensure adherence to principles of good fact-checking<sup>1</sup>

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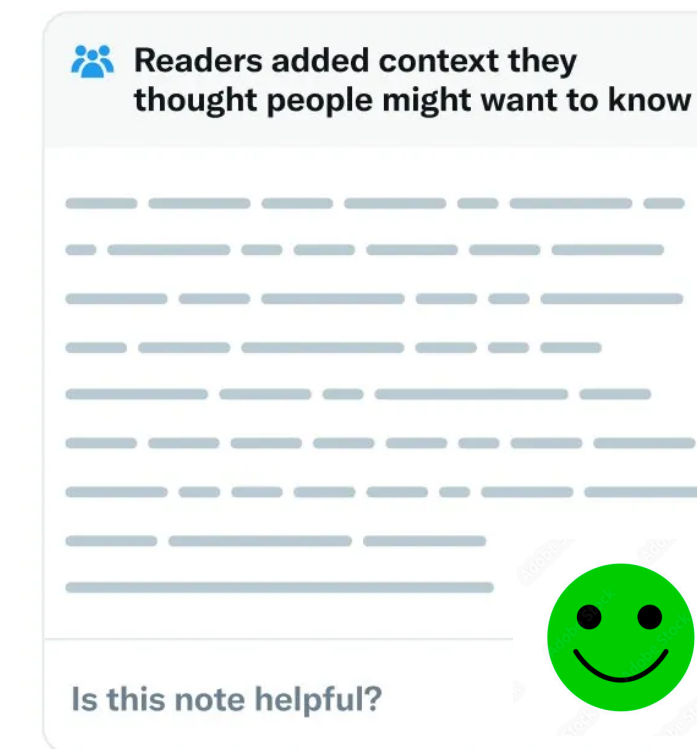
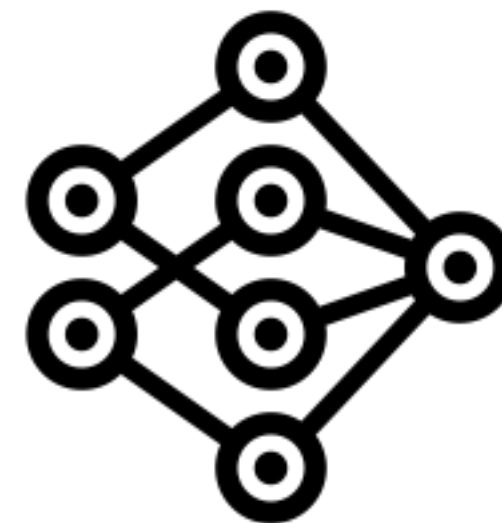
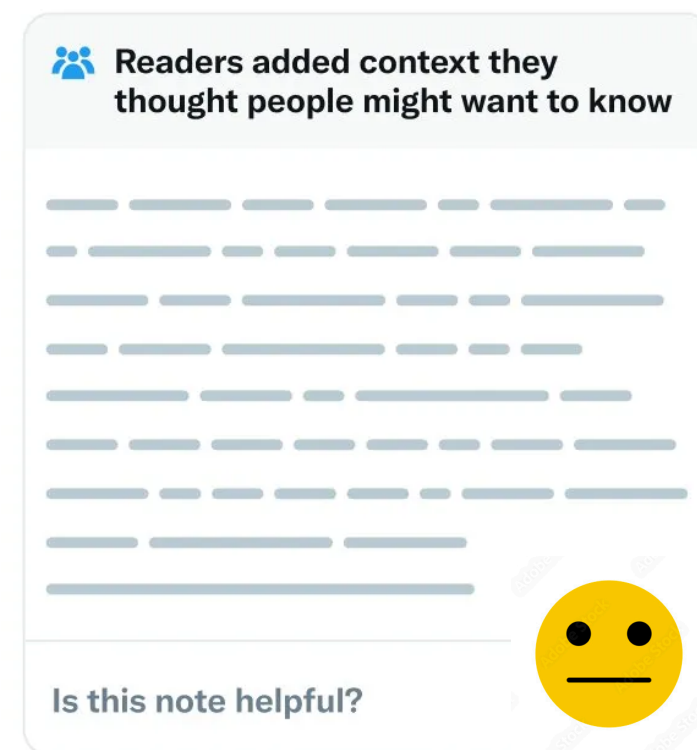
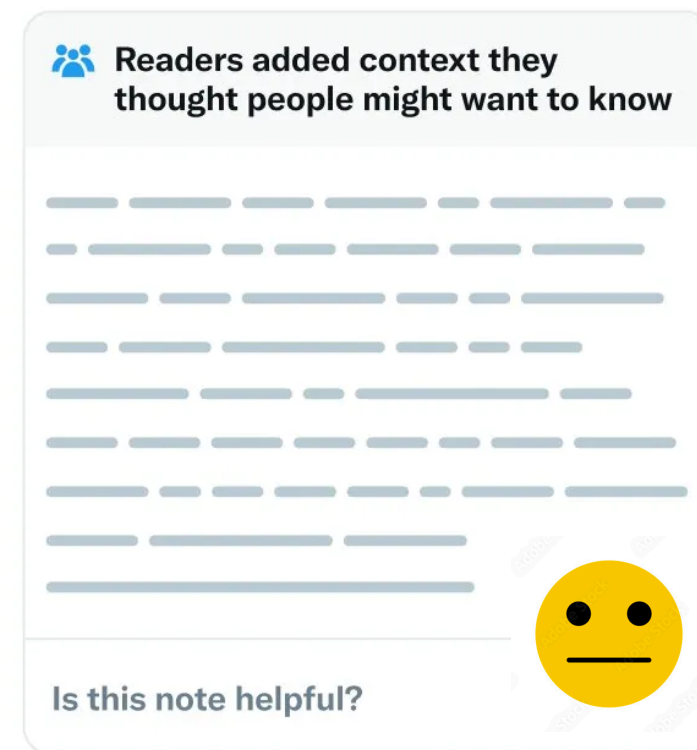
## 2. Scale up crowd-sourced fact-checking

- Use LLMs to summarize existing fact-checks
- Produce notes more likely to draw cross-ideological agreement



# Proposed Solution

## A Supernote

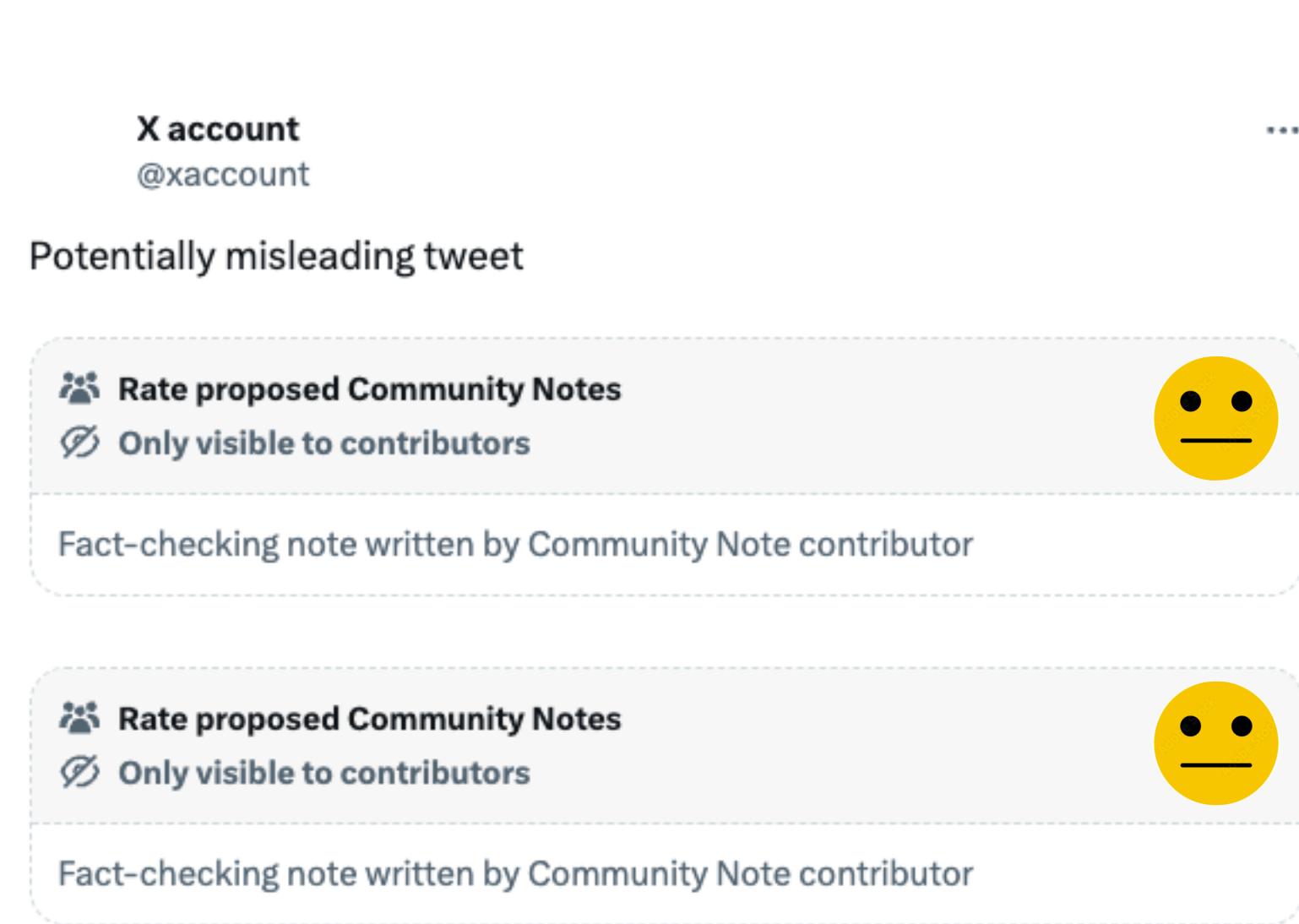


**An LLM-generated Super Note**

Take notes that **aren't rated Helpful yet**

# Proposed Solution

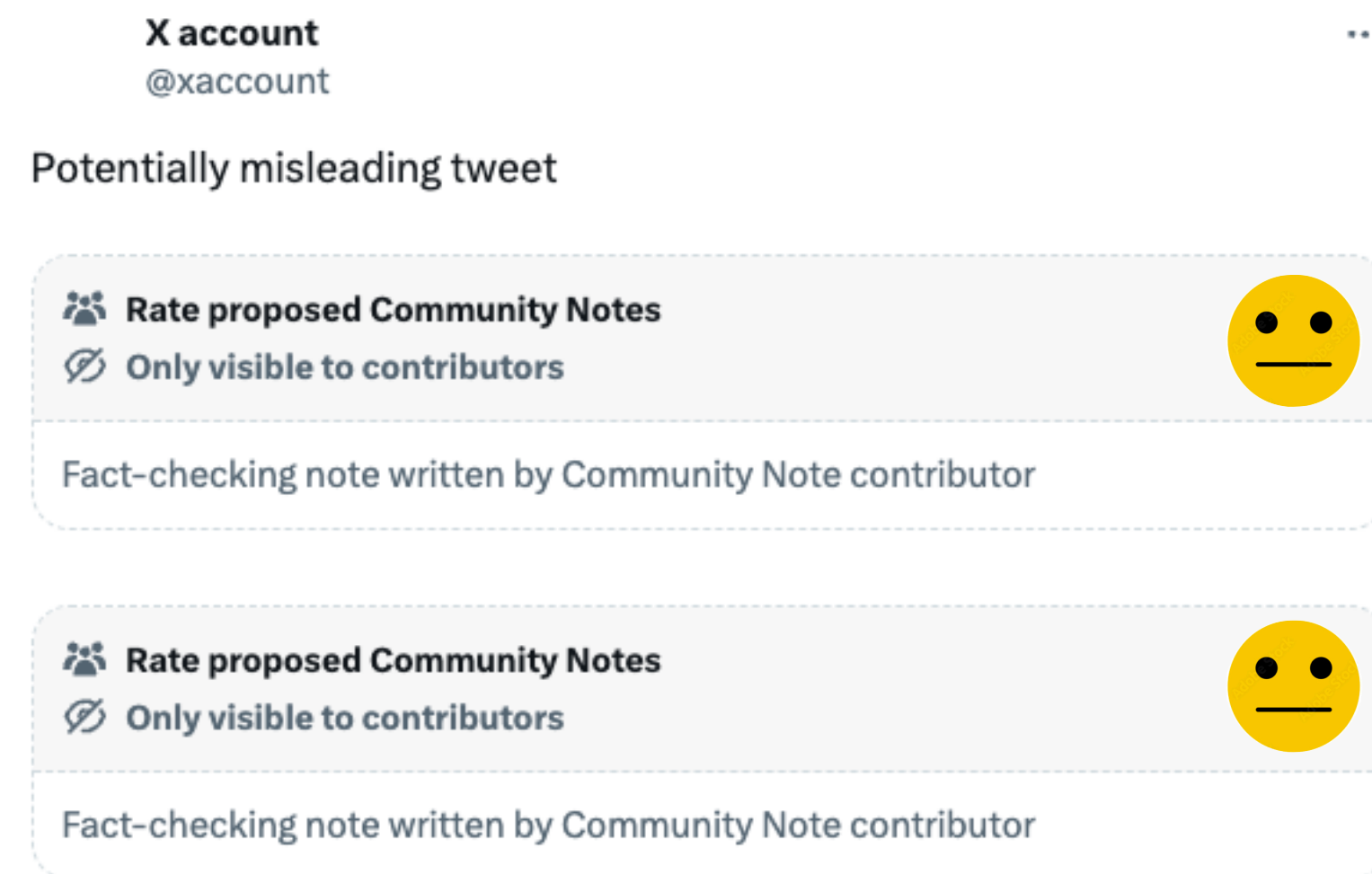
## In practice



This tweet has a few notes on it, but **none of them are currently rated helpful**

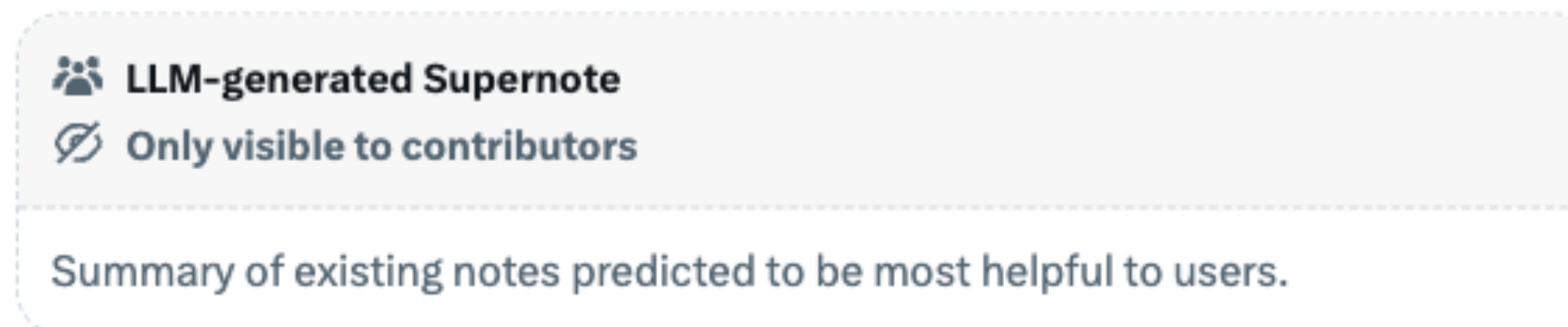
# Proposed Solution

## In practice



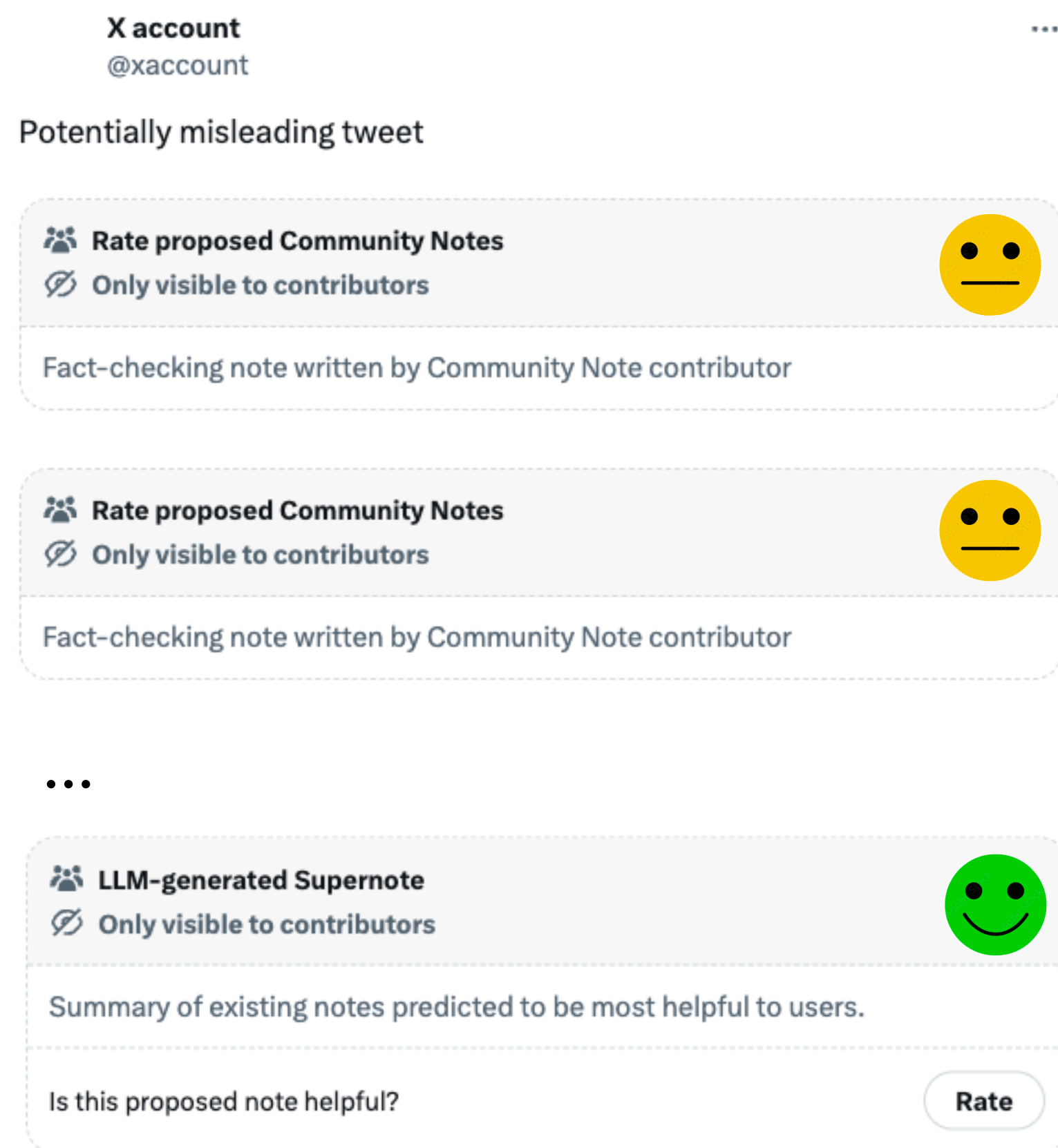
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# Proposed Solution

## In practice



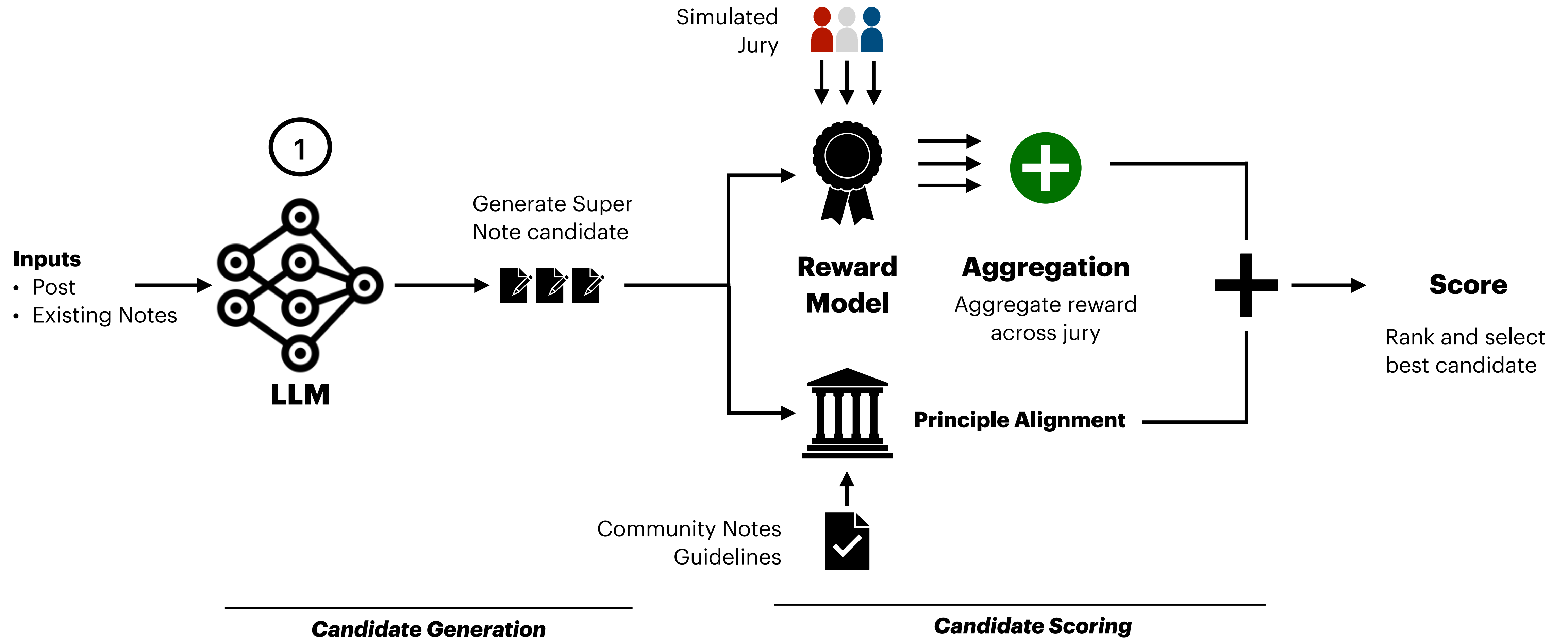
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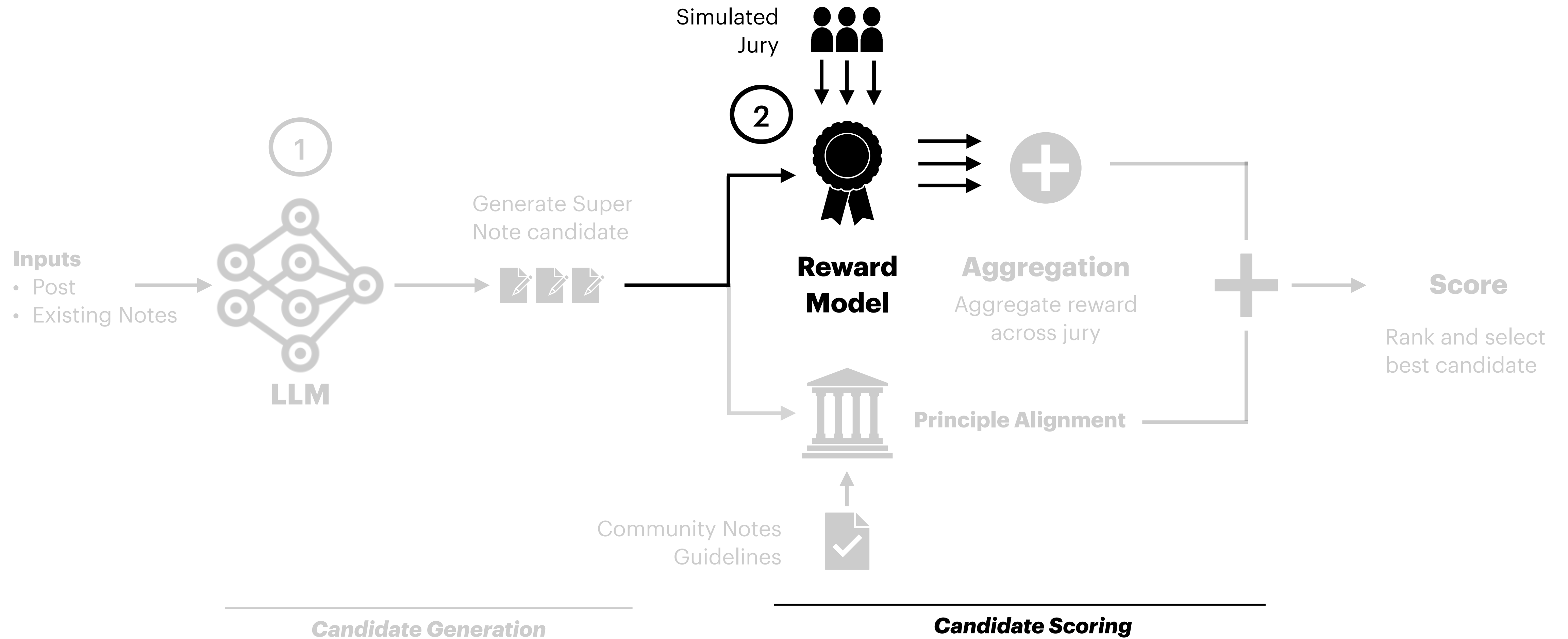
If the Supernote is expected to be **more helpful than any existing note**, it is shown on the platform alongside other candidate notes.

So, how to make a Supernote?

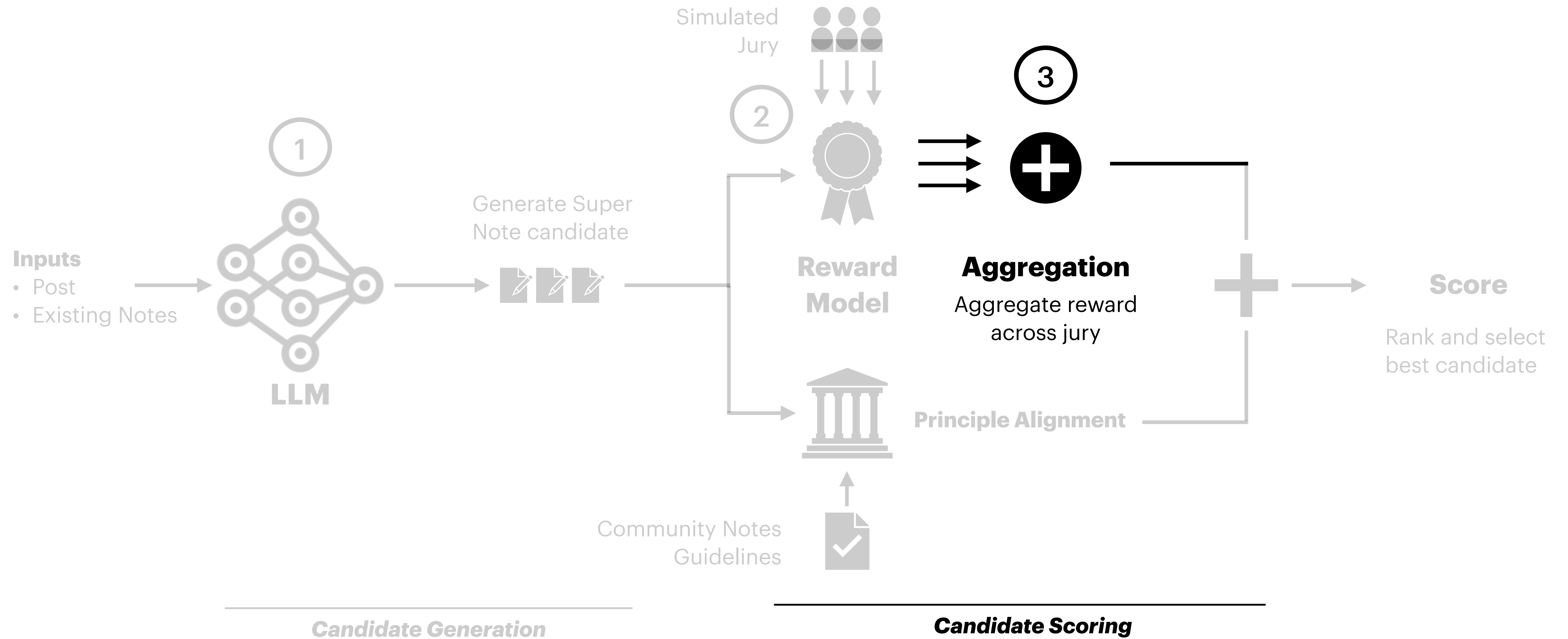
# Supernote Generation Pipeline



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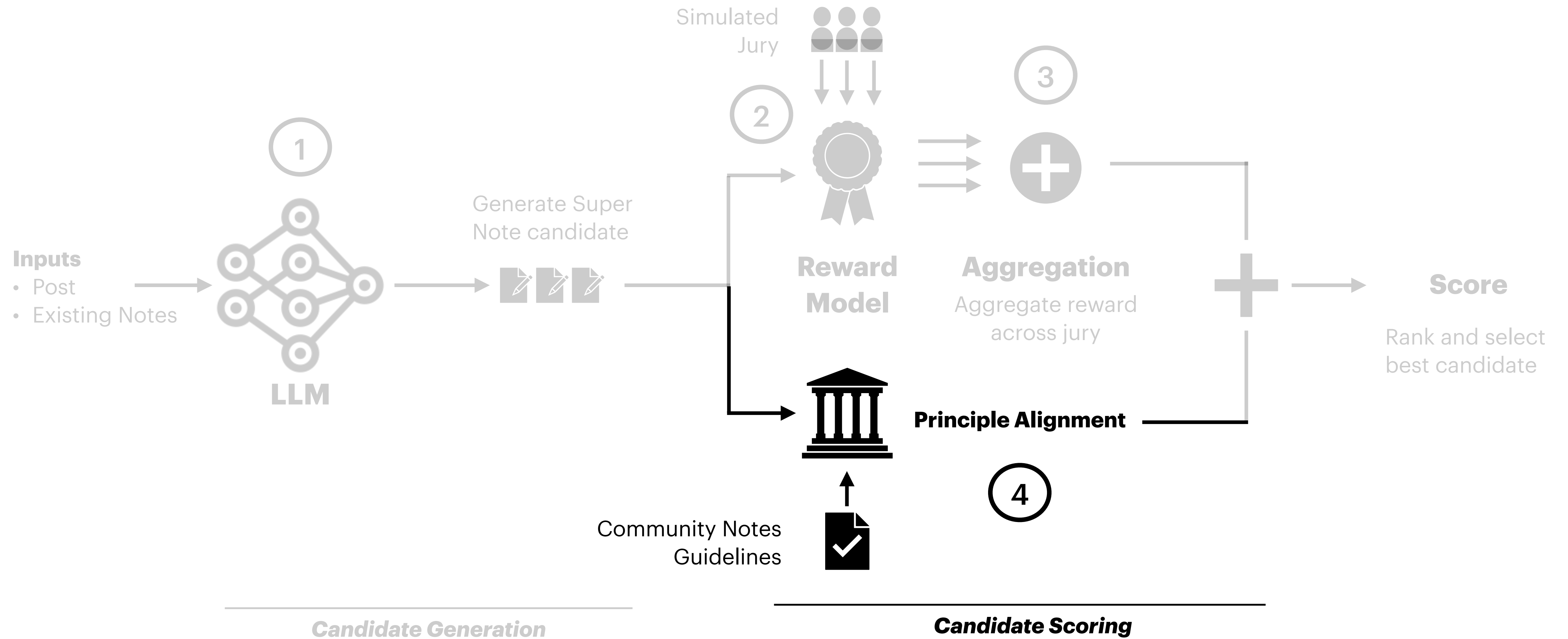


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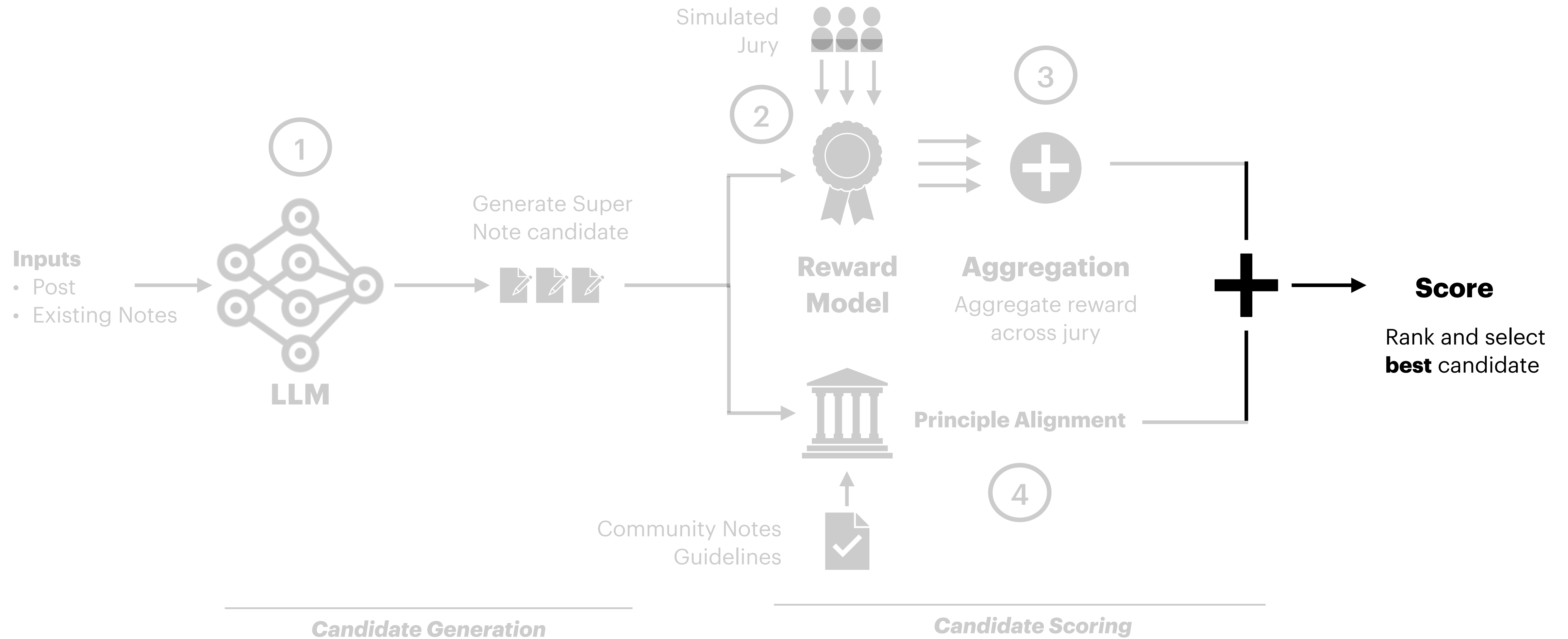




# Supernote Generation Pipeline



# Supernote Generation Pipeline



# 1 LLM Summarization

Generates 100 candidate supernotes by summarizing existing notes

X account  
@xaccount

Potentially misleading tweet

Rate proposed Community Notes  
Only visible to contributors

Fact-checking note written by Community Note contributor

Rate proposed Community Notes  
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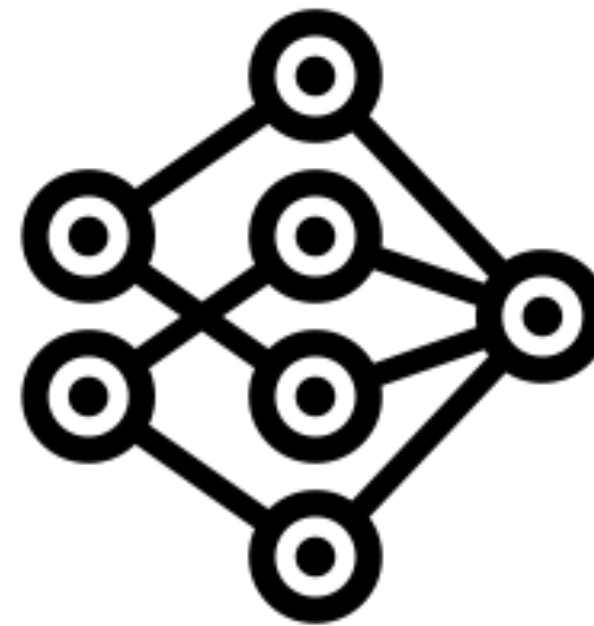
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**GPT 4**

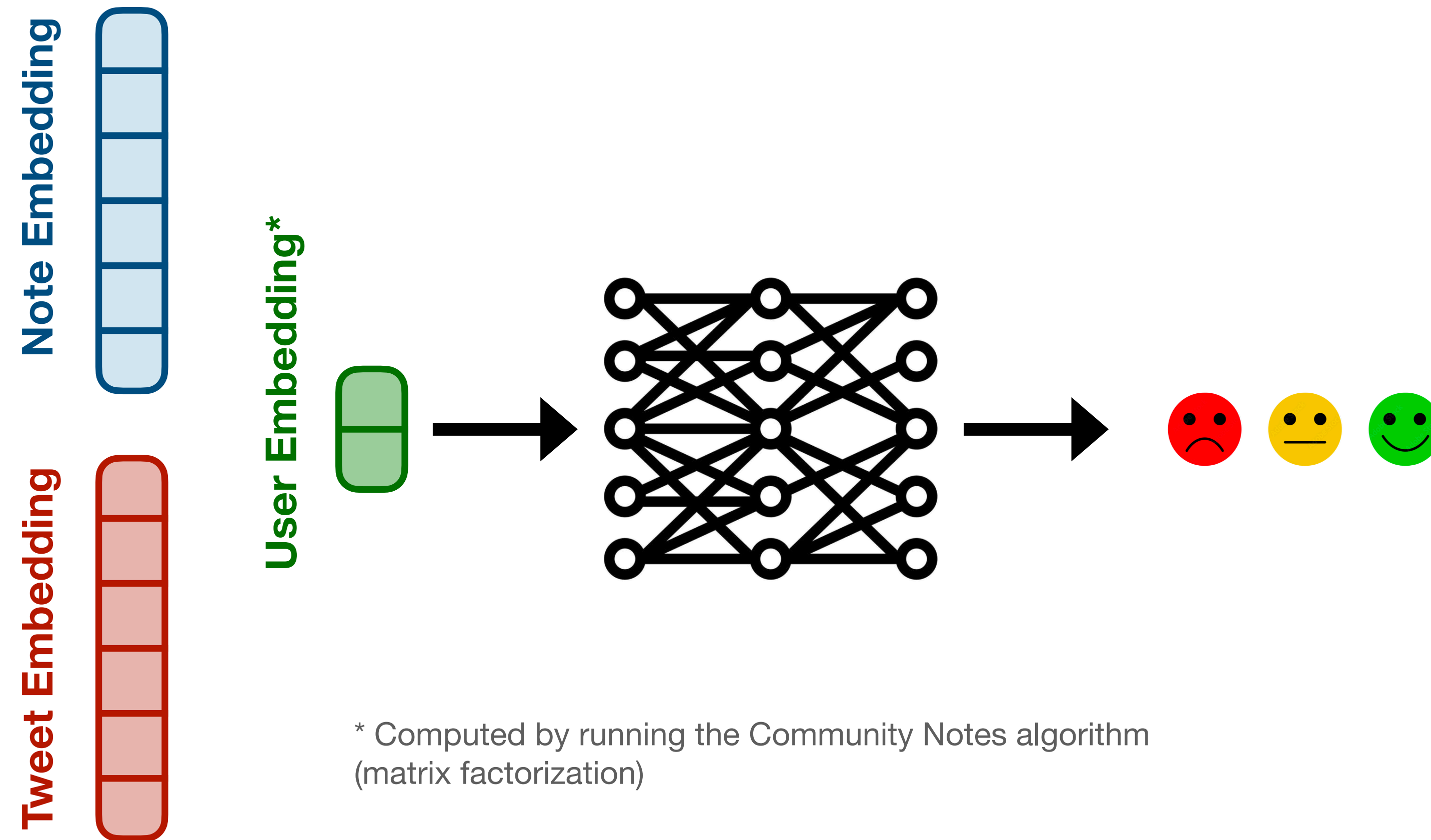


- Note Permutations
- temp=0.9
- top\_p=0.9

- **Candidate Super Note 1**
- **Candidate Super Note 2**
- ...
- **Candidate Super Note 100**

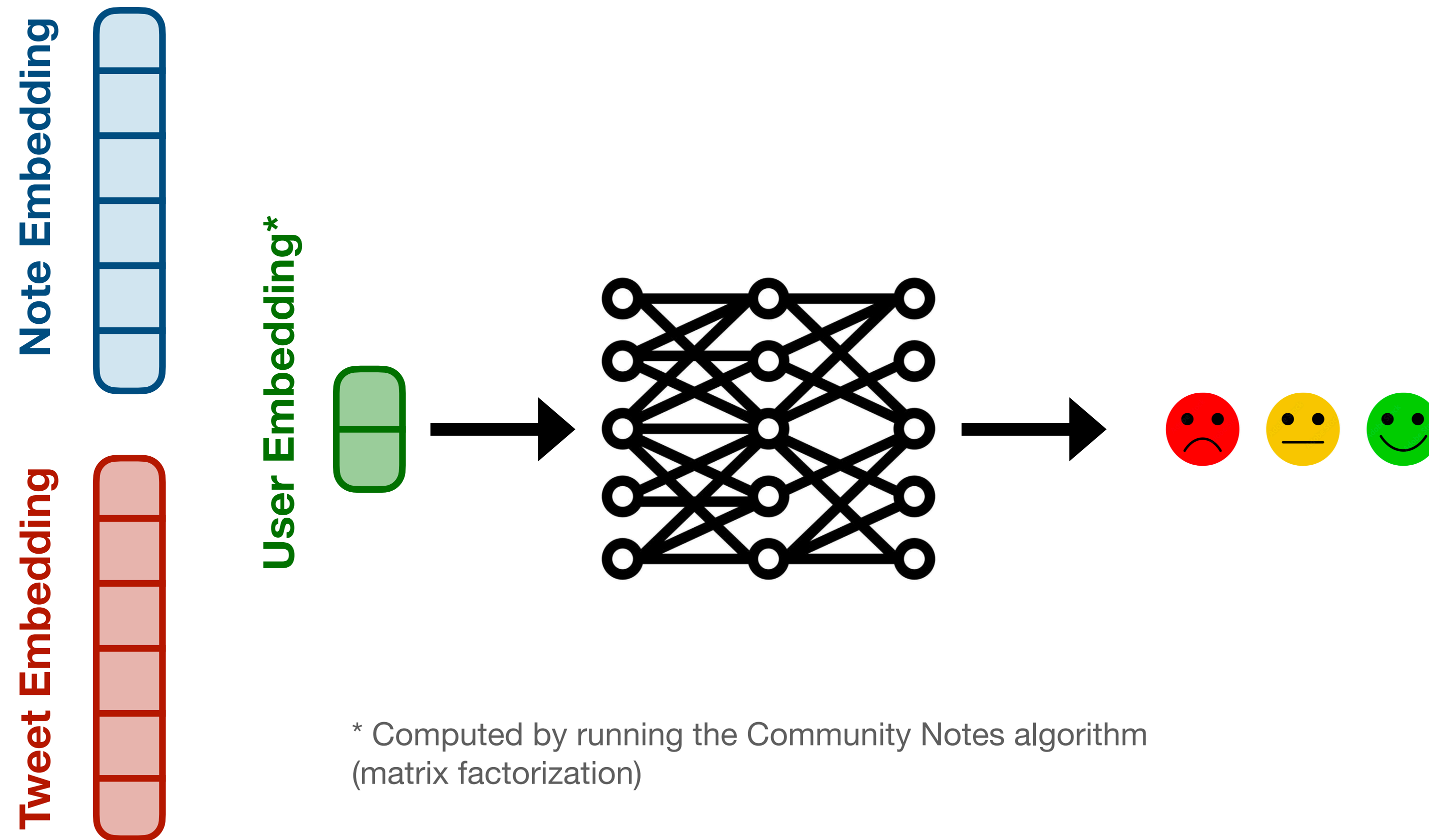
# 2 Reward Model

## Rater-conditioned helpfulness predictions

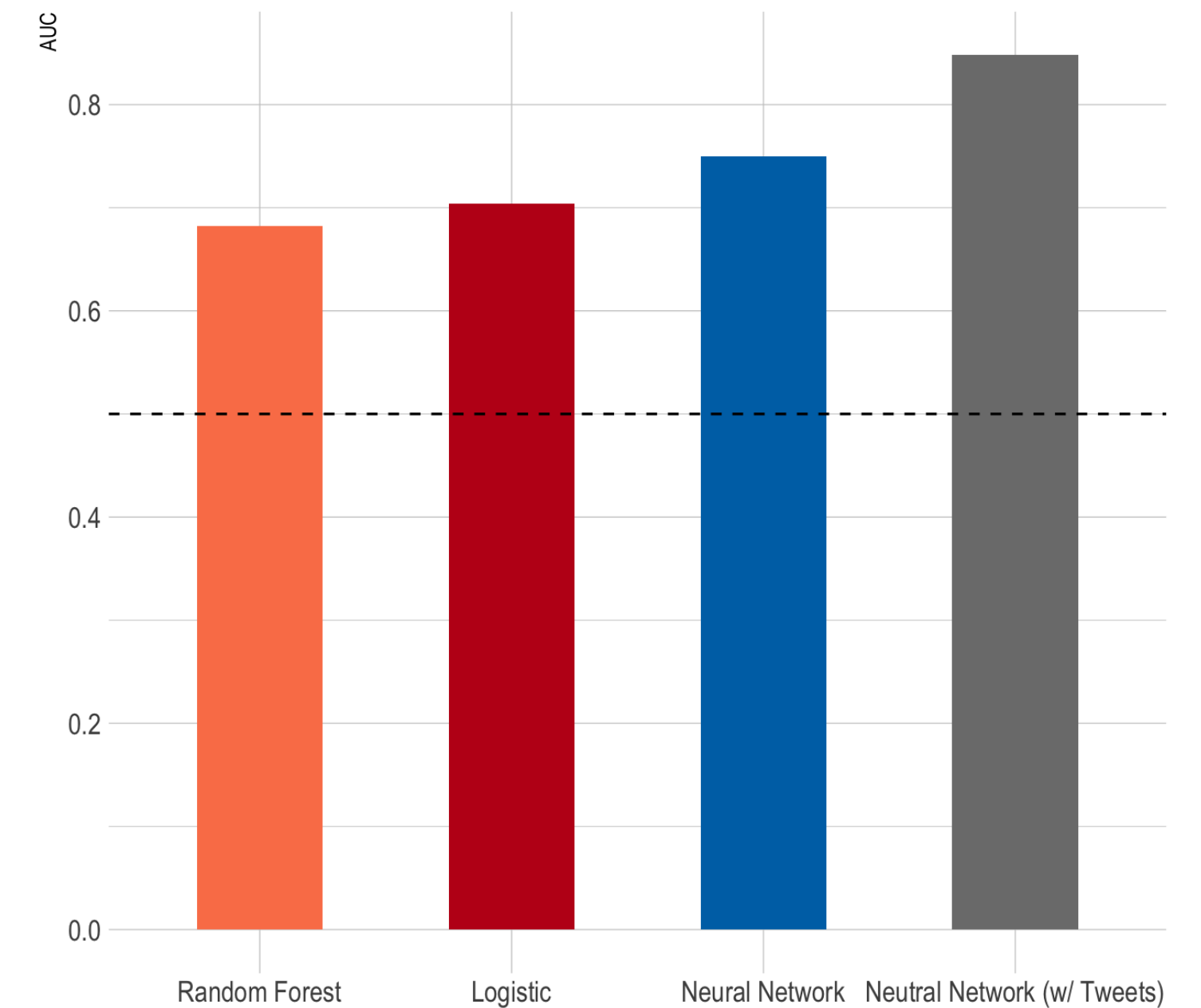


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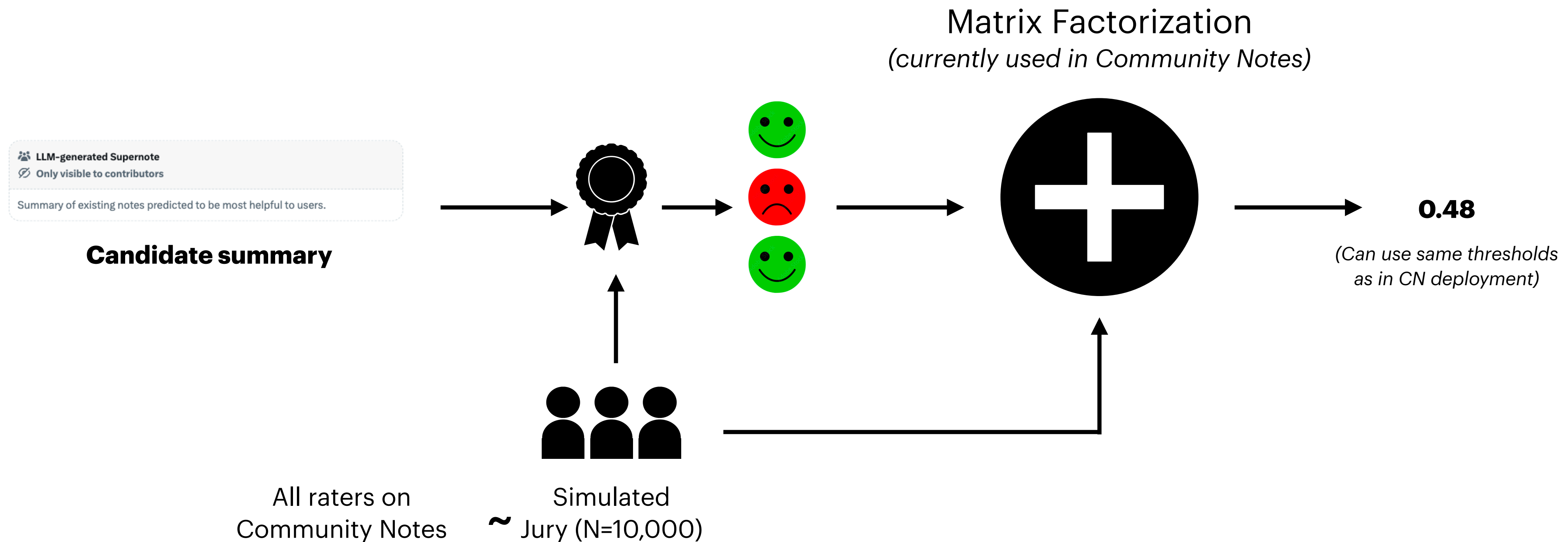


\* Computed by running the Community Notes algorithm (matrix factorization)



# 3 Aggregating Rewards

## Get a single score for a candidate summary



# 4 Principle Alignment

**Reject candidates that do not follow principles**

- Cites high-quality sources
- Easy to understand
- Directly addresses the post's claim
- Provides important context
- Neutral or unbiased language

Cool, does it work?



# Human Evaluation

We run a pilot (n=15) to evaluate Supernotes against existing notes

A

 Rate proposed Community Notes

 Only visible to contributors

Fact-checking note written by Community Note contributor

B

 LLM-generated Supernote

 Only visible to contributors

Summary of existing notes predicted to be most helpful to users.

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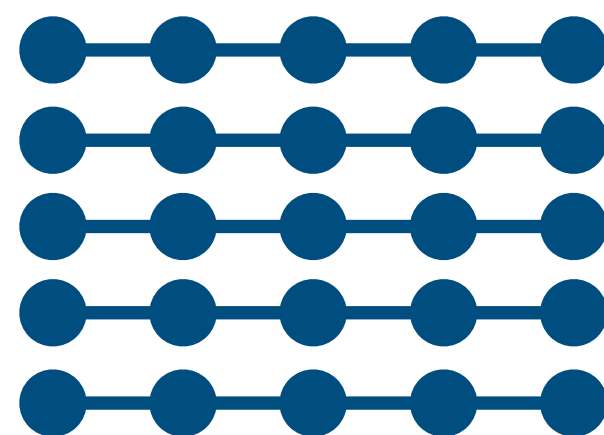
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1. Is this note helpful?



2. Agree/Disagree:

- a. High quality sources
- b. Clarity
- c. Comprehensive
- d. Context
- e. Non argumentative



B

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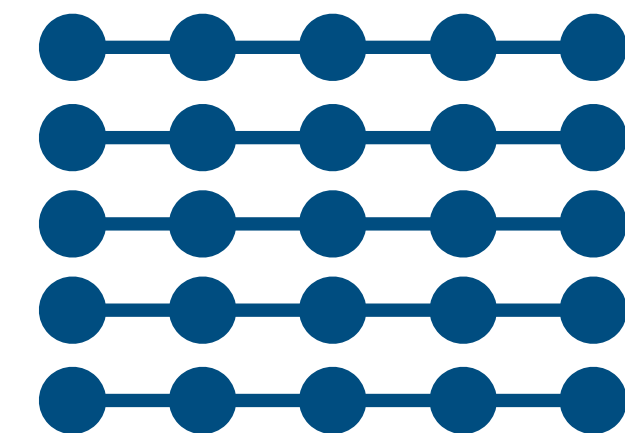
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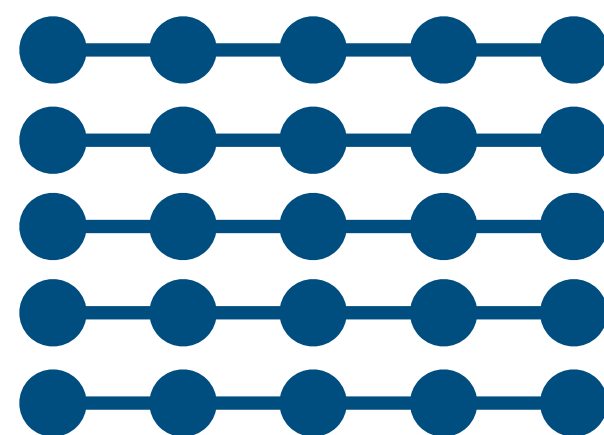
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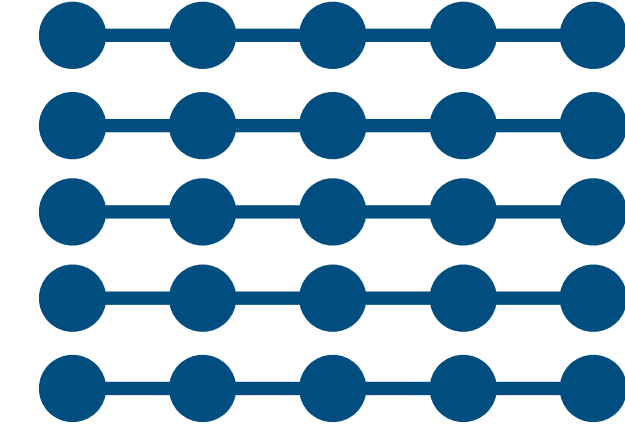
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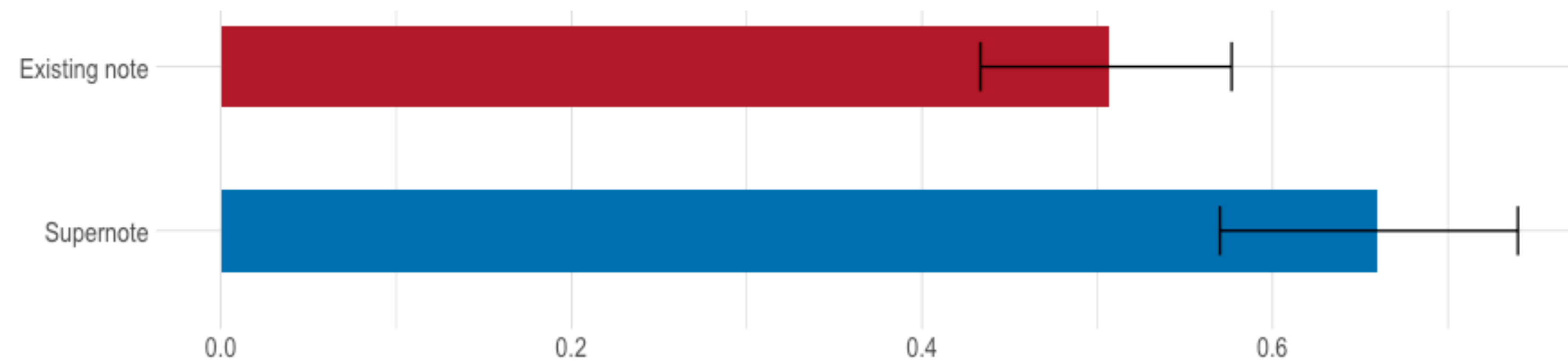


3. Which one is better?

A B

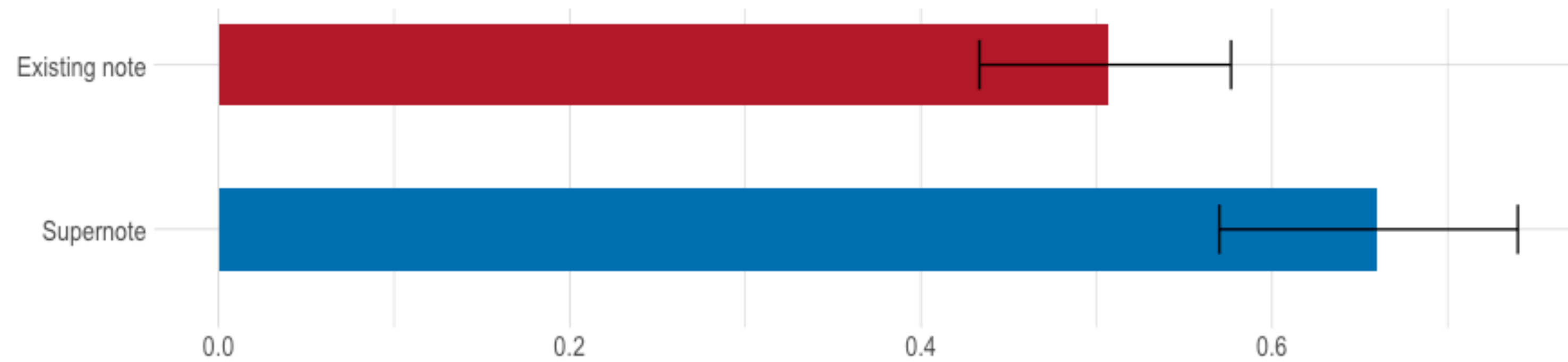
# Supernotes vs. Existing Notes

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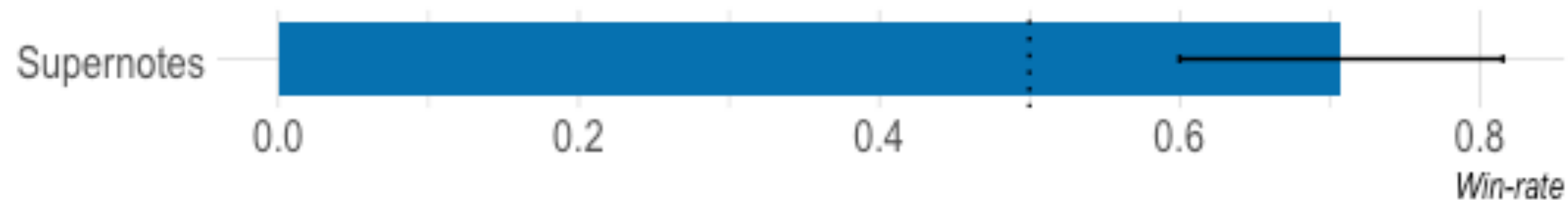


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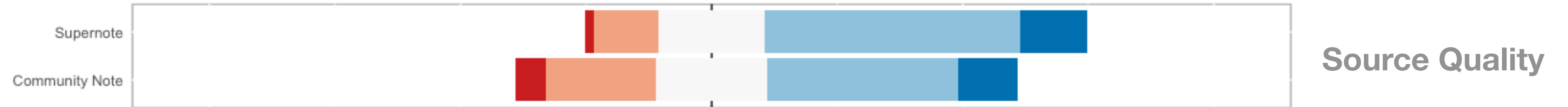


## 2. Even when an existing note is helpful, a Supernote is preferred



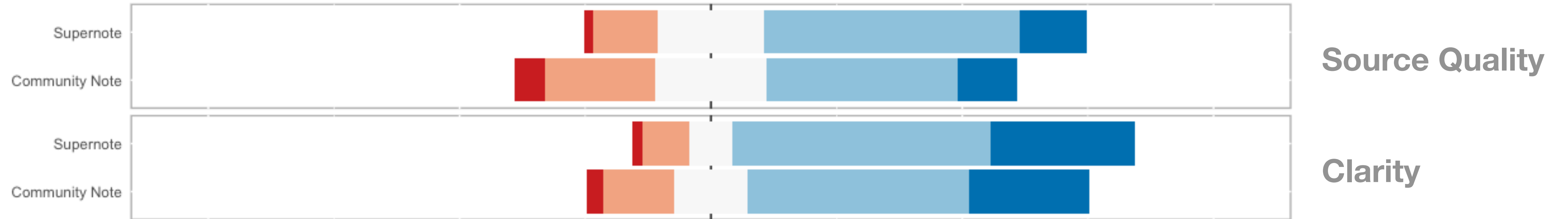
# Supernotes Attributes

## 3. They follow guidelines for good fact-checking



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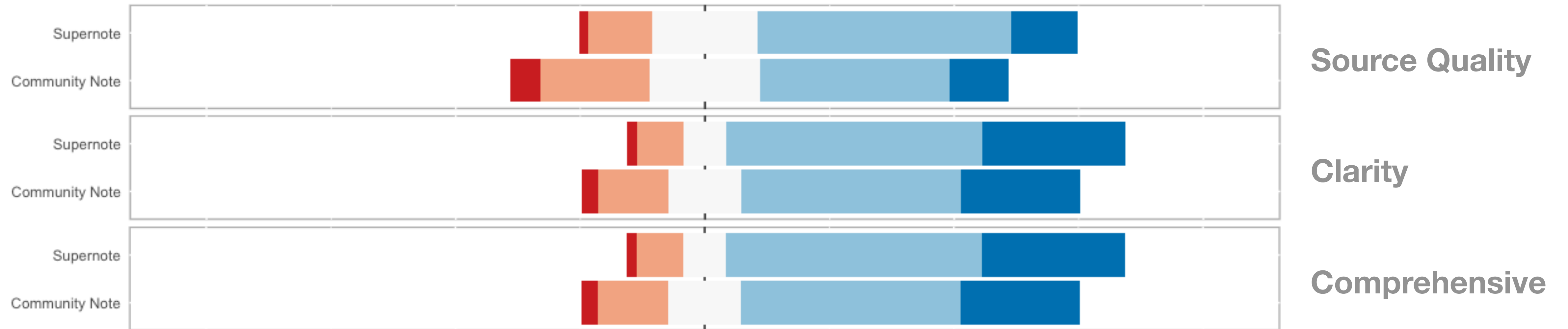
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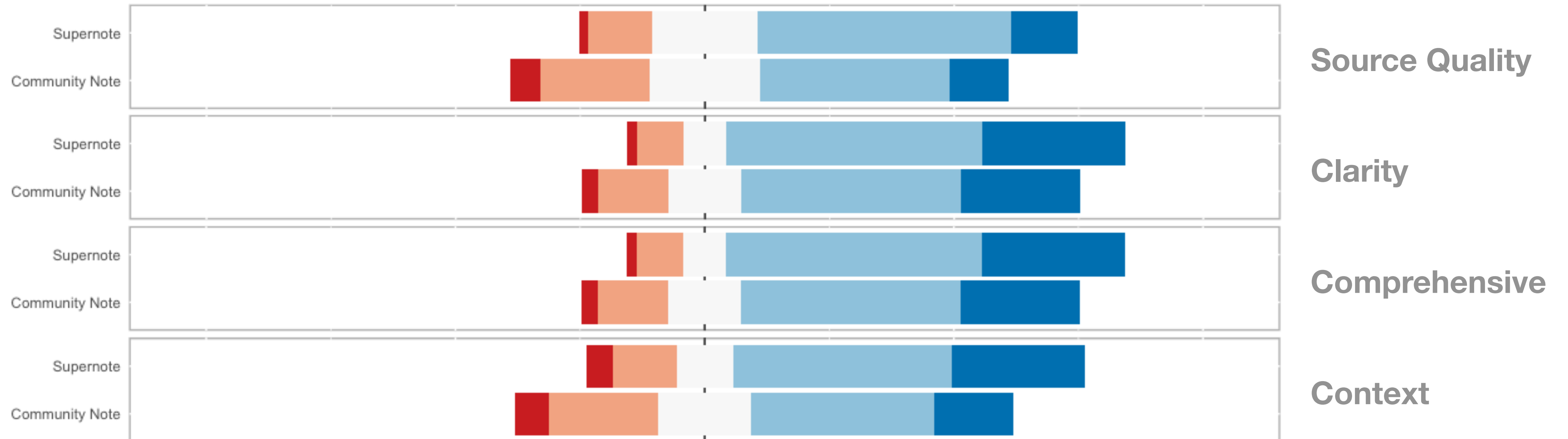
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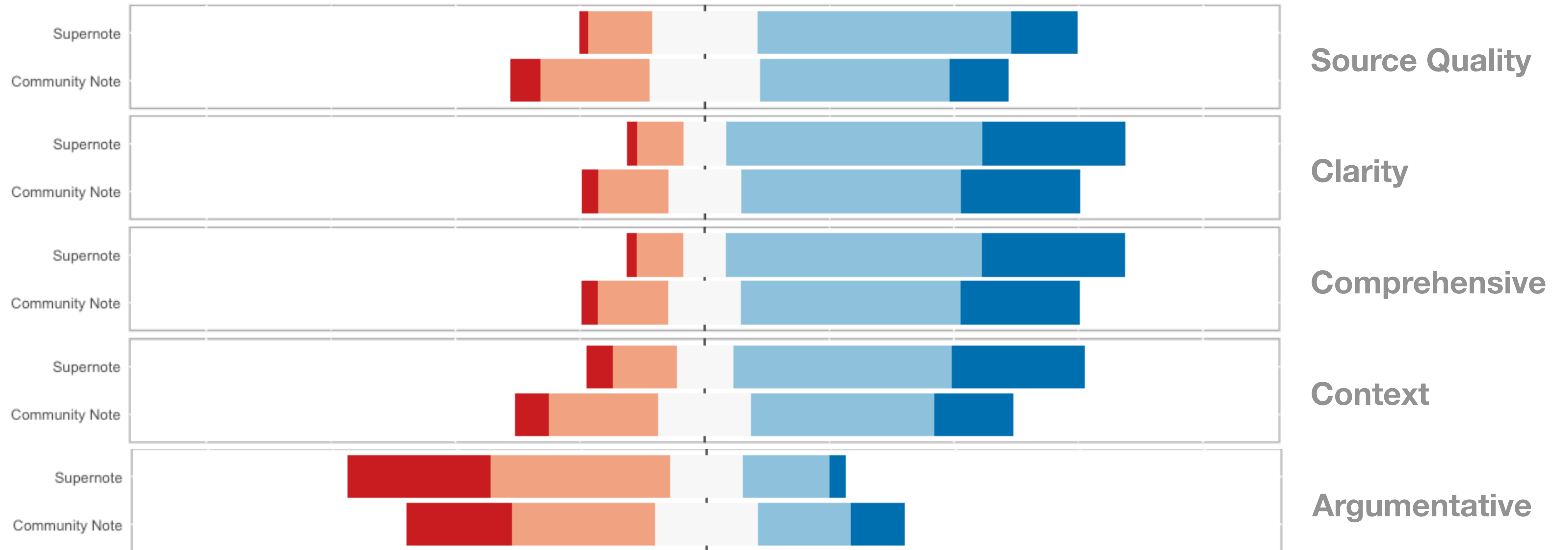
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# Ongoing Work

1. A larger study with more human participants
2. Ablation studies to measure impact of each part of our pipeline
3. Attribution issues - humans must get partial credit for writing notes that are used in a Supernote

# The *Supernote*

Enhancing Crowd-sourced factchecking using  
LLM-driven consensus

Soham De | [sohamde@uw.edu](mailto:sohamde@uw.edu)

IC<sup>2</sup>S<sup>2</sup> 2024

