

Post-hoc Study of Climate Microtargeting on Social Media Ads with LLMs: Thematic Insights and Fairness Evaluation

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Climate Change

- One of the most **urgent** challenges of the 21st century.
- Requires **broad public engagement** and **effective communication** to drive environmental action (*Moritz & Agudo, 2013; Dessler & Theater, 1995*).



Climate Debate in Social Media

- One of the most **urgent** challenges of the 21st century.
- Requires **broad public engagement** and **effective communication** to drive environmental action (*Moritz & Agudo, 2013; Dessler & Theater, 1995*).
- Social media has become a key arena for climate communication.
- Empowers organizations, activists, and policymakers to:
 - ▶ Disseminate information, Mobilize public support, Shape climate discourse (*Nosek, 2020; Hestres & Hopke, 2017; Adger et al., 2003*).



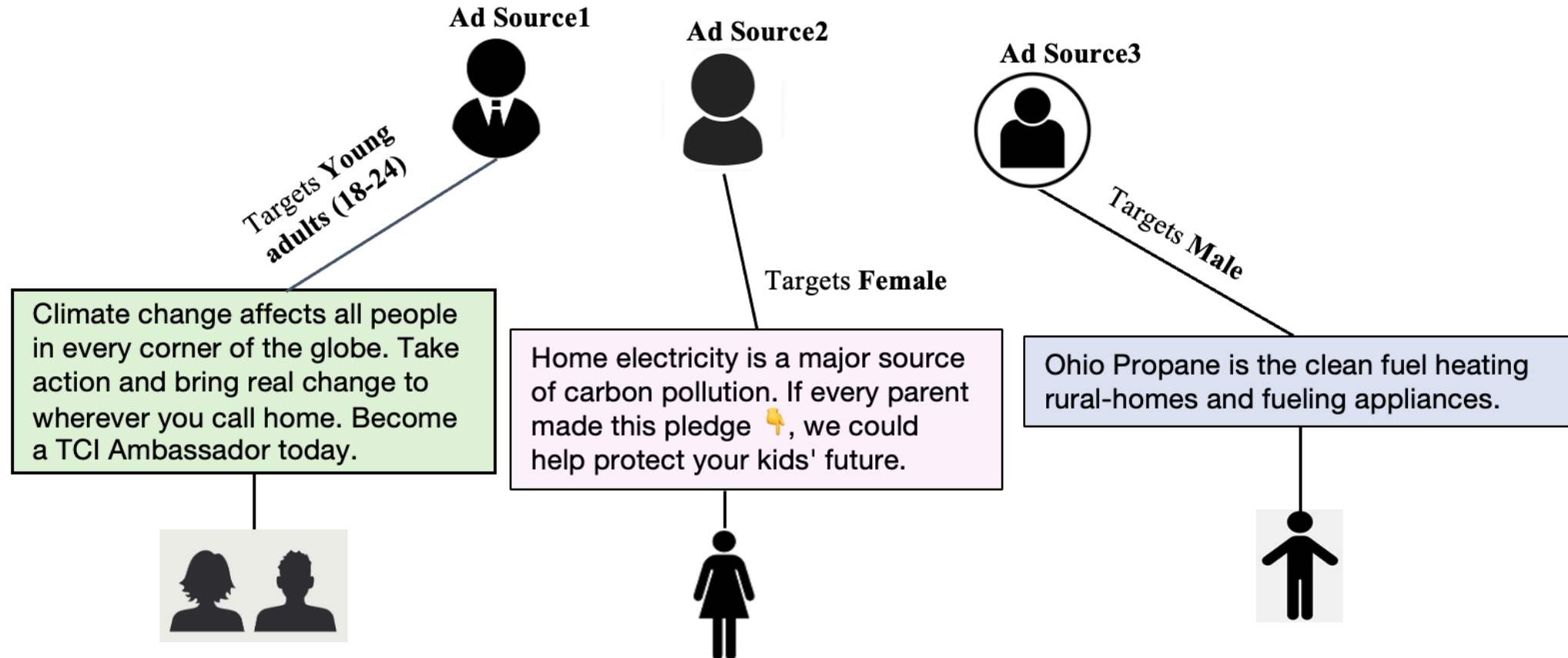
Microtargeting

Targeted Messaging Power

- Enables tailoring of messages to **specific demographics**.
- Increases message **relevance, engagement, and persuasive impact** (*Bloomfield & Tillery, 2019; Walter et al., 2018; Stoddart et al., 2016*).
- Risk reinforcing bias, exclusion, and inequity — raising critical questions about **fairness and transparency** in digital climate communication.

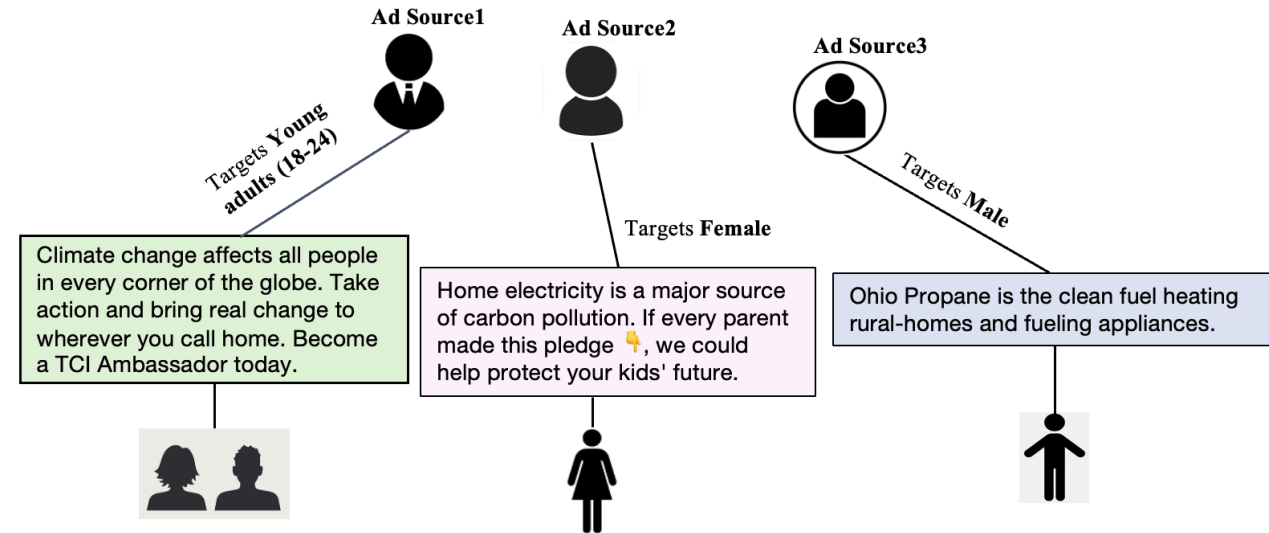


Microtargeting in Climate Debate



Microtargeting in Climate Debate

- **Problem:** Limited understanding of *how* microtargeting strategies are crafted and whether they're **fair or biased**.
- **Goal:** Leverage LLMs to analyze microtargeting and evaluate **demographic targeting** and **fairness**.



Research Questions (RQ)

- **RQ1:** Can LLMs **identify** targeted demographics and **explain** their reasoning?
- **RQ2:** What are the recurring themes and aspects of **explanations** provided by LLMs?
- **RQ3:** How **fair** are LLM predictions across demographic groups?

Dataset: Climate Campaigns Case Study

- Source
 - Based on Islam et al. (2023b) and Islam & Goldwasser (2024a)
 - **Corpus:** 21,372 English climate-related Meta ads (U.S.)
 - **Time span:** Jan 2021 – Jan 2022
- Ad Attributes:
 - Ad ID, description, body, funding entity, spend, impressions
 - Impression breakdowns by: **Gender, Age, Location:** U.S. state level
- **Demographics Indicators:**
 - **Gender:**
 - *Male,*
 - *Female*
 - **Age groups:**
 - *Young adults 18-24,*
 - *Early working 25-44,*
 - *Late working 45-64,*
 - *Seniors 65+*

Dataset: Climate Campaigns Case Study

- Targeting Definition
 - Meta API does **not** provide explicit targeting
 - Targeting **inferred** from *exclusive impression* distributions
 - Ads viewed **only by one gender** or **only by one age group**
 - Ads overlapping multiple categories **excluded**
- Filtered Dataset

Category	Targeting Basis	# Ads
Gender-targeted	Male-only / Female-only	106
Age-targeted	Exclusive to one age group	121
Total (after filtering)	—	227

Experimental Setup

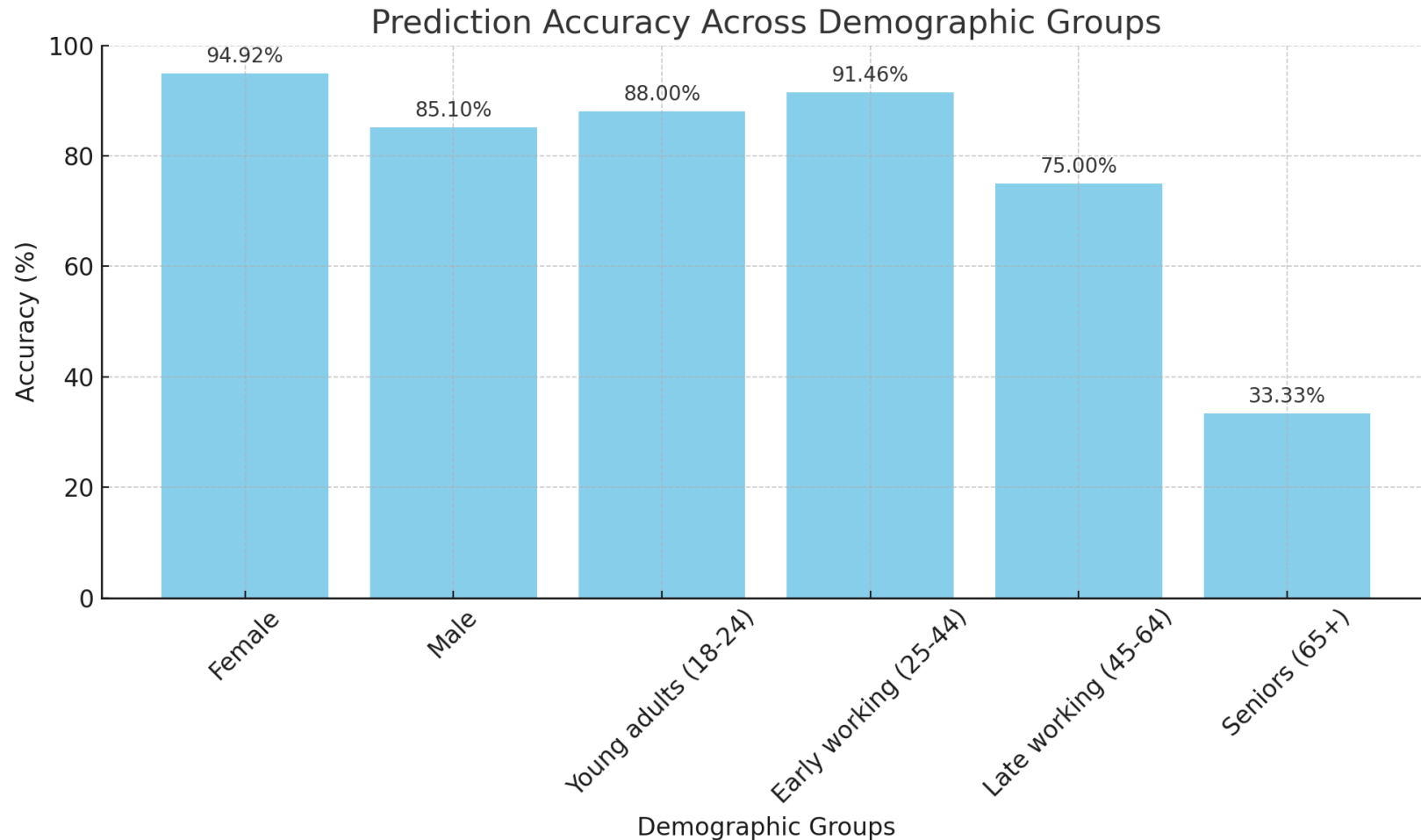
- **Task: Identify** the targeted demographic in a text and provide an **explanation**.
- Pipeline steps
 - **Gender prediction with explanation:** Given a text, predict the targeted gender and justify the choice.
 - **Age group prediction with explanation:** Given a text, predict the targeted age group and justify the choice.

Experimental Setup

- **Tools used:** OpenAI's latest LLM (**o1-preview**)
- **Baseline comparisons:** LR_{TF-IDF}, BERT, Llama 3, Mistral Large 2
- **Fairness metrics:**
 - ▶ Demographic Parity,
 - ▶ Equal Opportunity,
 - ▶ Predictive Equality.

Results

- Prediction Accuracy Across Demographic Groups



Results

Model	Demo.	Acc. (%)	F1 (%)
LR_tf-idf	gender	69.00	65.00
LR_tf-idf	age	73.00	31.00
BERT	gender	72.00	70.00
BERT	age	70.00	26.00
Llama 3	gender	80.19	79.67
Llama 3	age	58.68	36.84
Mistral Large 2	gender	82.08	82.07
Mistral Large 2	age	74.38	48.68
o1-preview	gender	90.57	90.35
o1-preview	age	85.95	71.00

Table 1: Baseline Comparisons.

Results

Gender	p-value	Conclusion
Male	5.35×10^{-7}	Reject H_0
Female	5.95×10^{-14}	Reject H_0

Table 2: Significance tests by subgroup (gender).

Age Group	p-value	Conclusion
Senior	4.66×10^{-1}	Fail Reject H_0
EarlyWorking	1.04×10^{-35}	Reject H_0
Young	5.76×10^{-11}	Reject H_0
Late Working	4.23×10^{-3}	Reject H_0

Table 3: Significance tests by subgroup (age).

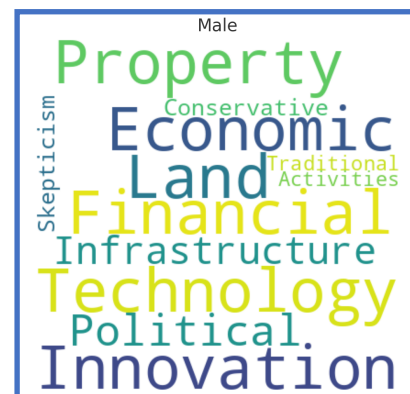
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Table 1: Baseline Comparisons.

Thematic Insights (Explanations from LLM)



perceived interests and roles in technology, finance, property, traditional male activities, and political or economic discourse

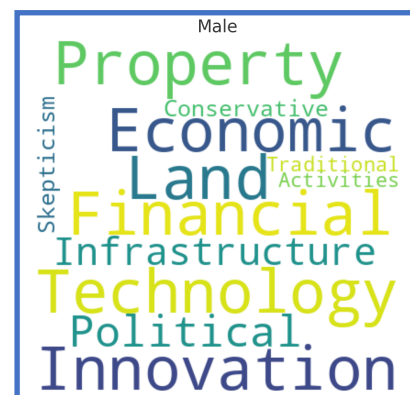


Thematic Insights (Explanations from LLM)



Roles as Caregivers,
Environmental Advocates, and
Socially Conscious Individuals

perceived interests and roles in technology,
finance, property, traditional male activities,
and political or economic discourse



Thematic Insights (Explanations from LLM)

Male

Property
Economic
Land
Financial
Infrastructure
Technology
Political
Innovation

Skeptical
Conservative
Traditional
Activities

Early Working

Parental
Financial
Innovation
Stability
Career
Proactive
Technology
Responsible
Social

Female

Leadership
Community
Environmental
Welfare
Advocacy
Empathy
Social
Caregiving
Emotional

Health
Safety
Empowerment

Late Working

Policy
Financial
Stability
Economic
Homeownership
Responsibility
Engagement

Voter

Young Adults

Activism
Bold
Passion
Leadership
Change
Advocacy
Action
Environmental
Immediate
Climate
Training

Seniors

Health
Safety
Vulnerability
Wellness



Activism and Environmental
Consciousness

Thematic Insights (Explanations from LLM)



Proactive and Responsible
Mindset



Activism and Environmental
Consciousness

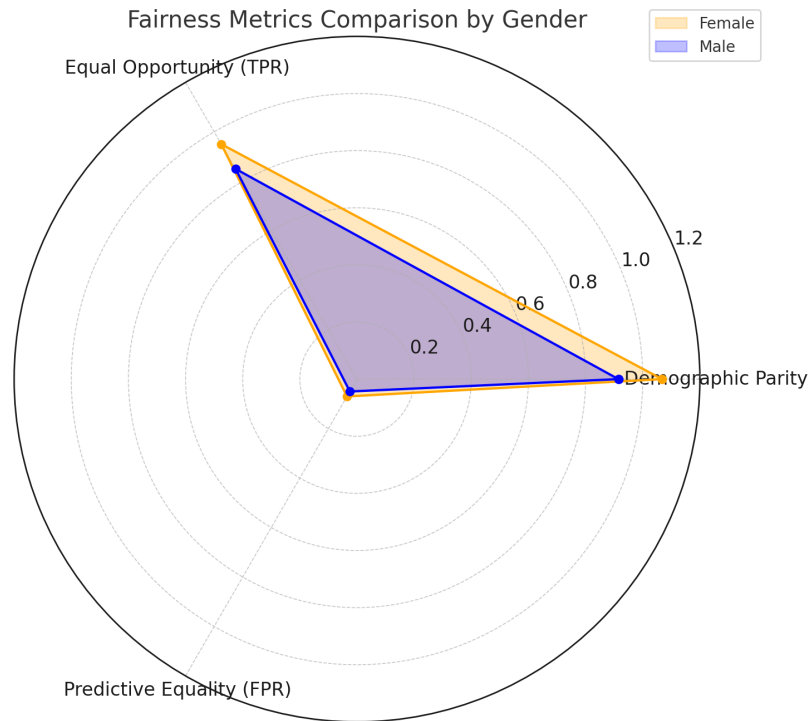
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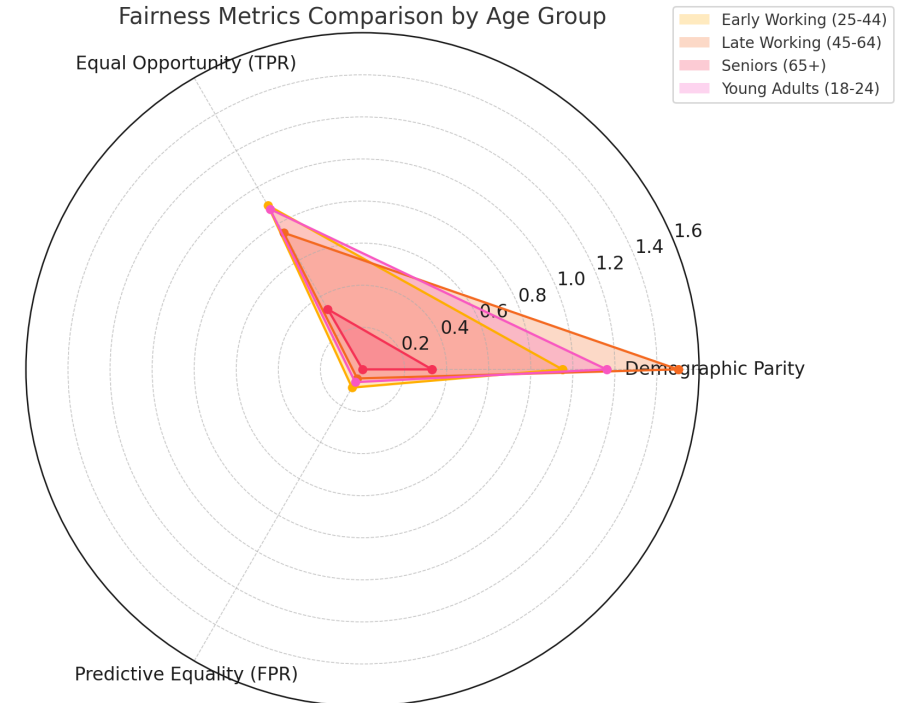
Thematic Insights (Explanations from LLM)



Fairness Evaluation



Minor bias favoring females.



- Minor bias favoring late working age group.
- Significant bias against seniors.

Key Takeaways

LLMs as Auditors

- LLMs can act as **independent, third-party auditors** of **microtargeted climate ads**.
- They can **identify intended audiences, explain reasoning, and reveal fairness issues** that are otherwise opaque.

Strong Predictive Capability

- **Gender prediction:** Highly accurate (Female = 94.92%, Male = 85.10%).
- **Age prediction:** More variable — highlights complexity of age-based targeting.

Key Takeaways

Explainable Insights

- LLM-generated **explanations** support creation of a **reusable taxonomy** of thematic appeals used in ads.
- These explanations make demographic targeting **interpretable and actionable** for researchers and practitioners.

Fairness & Equity

- **Fairness analysis** reveals **disparities** across gender and age groups.
- Demonstrates the **need for inclusive, equitable ad-targeting** and **bias-aware AI evaluation frameworks**.

References

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THANK YOU 😊

Slide: <https://tunazislam.github.io/files/PosthocClimateLLM.pdf>



Questions?

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