



# ChronoBias: A Benchmark for Evaluating Time-conditional Group Bias in the Time-sensitive Knowledge of Large Language Models

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## TL;DR

We propose a novel benchmark, ChronoBias, to highlight the importance of **time-conditional (temporal) group bias** analysis in fully uncovering the mechanisms of group bias in the **time-sensitive** knowledge of large language models (LLMs).

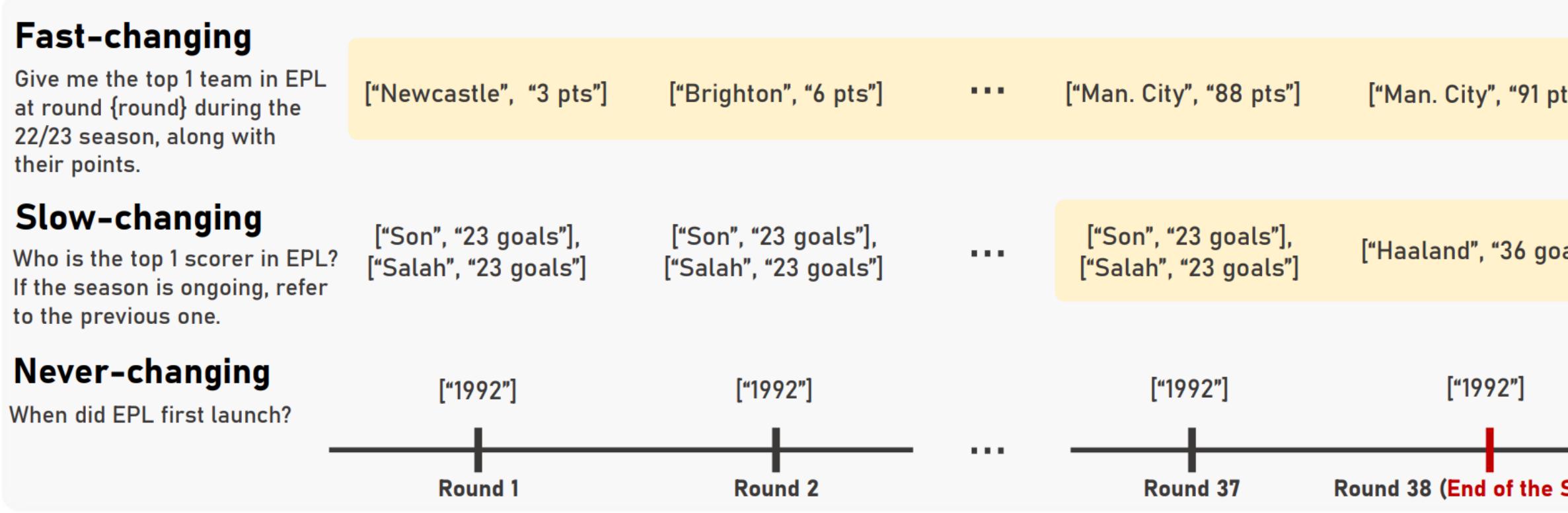
### Problem 1: Group Bias of LLMs

- LLMs often show disparate performance across factors such as demographic groups, regions, or levels of popularity.
- This undermines LLMs' fair deployment for diverse groups of users.

Metric	Brazil	Spain	England	Korea	Saudi	Algeria
Model Performance	0.378 (1)	0.275 (2)	0.237 (3)	0.14 (4)	0.136 (5)	0.073 (6)
Monthly Page-views	47,028 (3)	219,607 (2)	622,919 (1)	10,324 (5)	28,197 (4)	4,098 (6)

### Problem 2: Time-sensitivity of Knowledge

- LLMs must adapt to **time-sensitive** knowledge that are updated after their knowledge cutoffs, to maintain performance uniformly across all time intervals.
- Knowledge cutoff of an LLM: The most recent date of the knowledge used for its pretraining



### Possible Solution: Retrieval-augmented Generation (RAG)

- RAG can be a possible solution to aforementioned problems, since it...
  - supplements knowledge acquired before the knowledge cutoff that was not well retained in model parameters (**Problem 1**);
  - processes new information without further fine-tuning (**Problem 2**).
- Main Goal:** Construct RAG-based LLMs, which are fair and accurate uniformly across all time intervals!

### Research Questions (RQ)

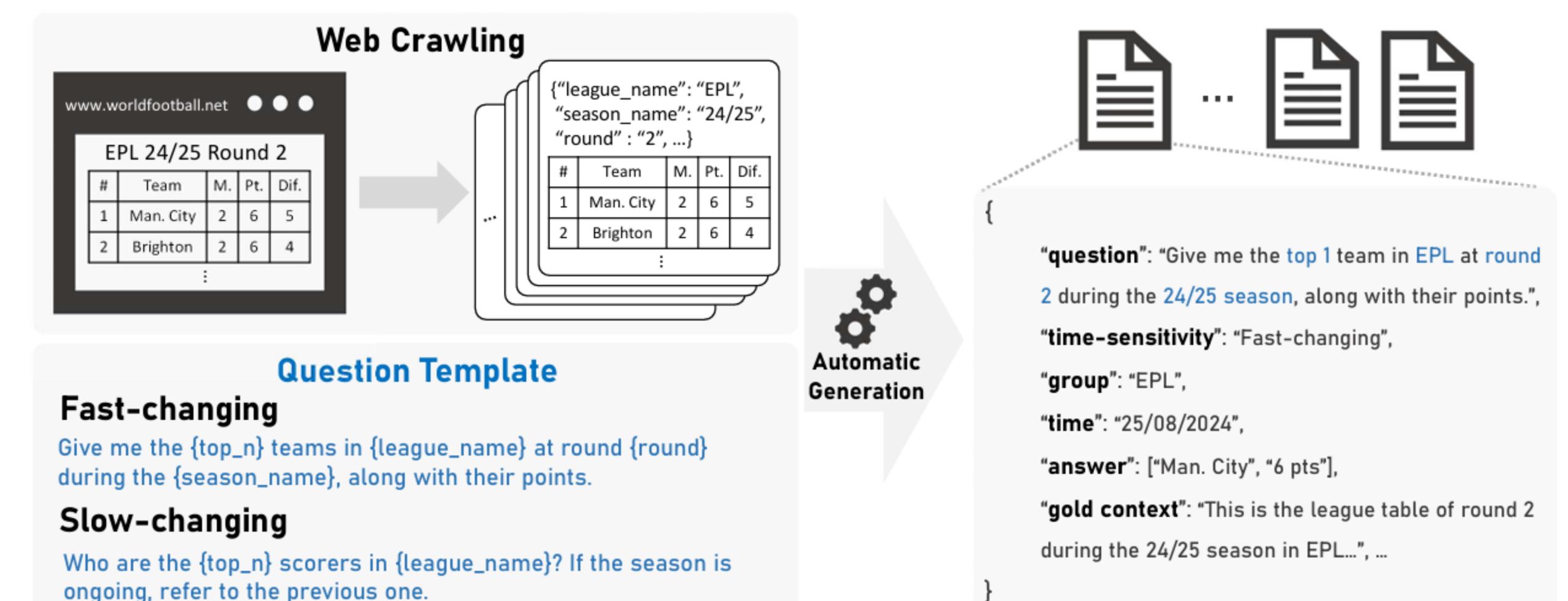
- RQ 1. Group bias in vanilla LLMs
- RQ 2. Rationale for group bias in vanilla LLMs
- RQ 3. Guidelines for building fair and accurate RAG-based LLMs

### Main Challenge: Lack of Group Bias Analysis in the Time-sensitive Knowledge of LLMs

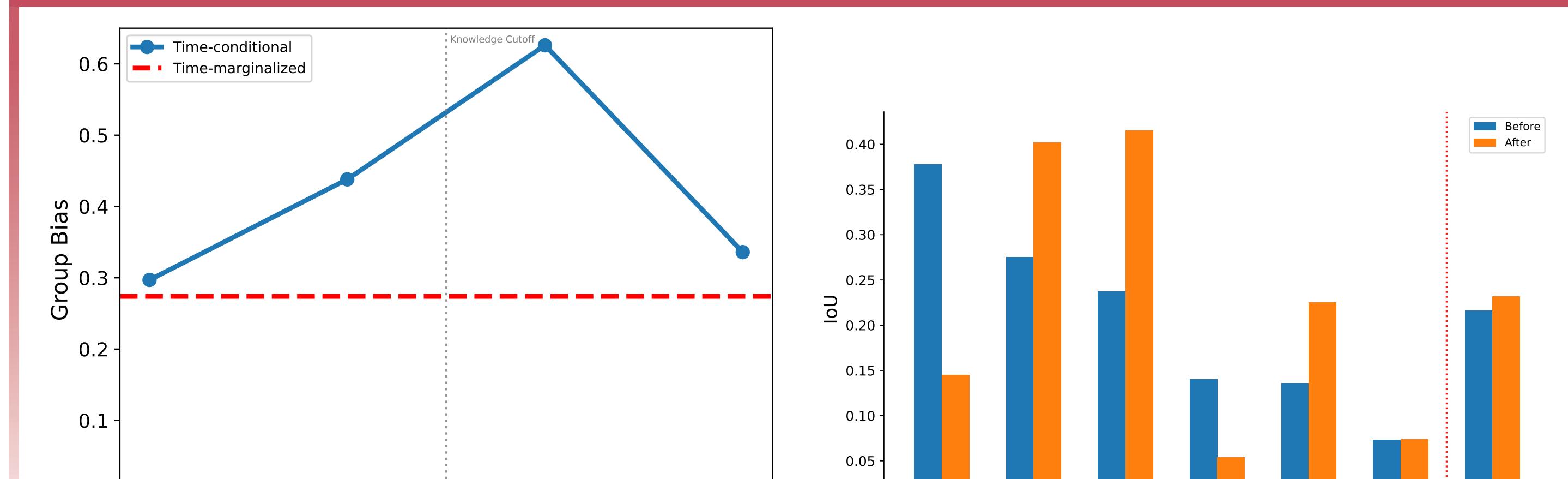
- For questions involving **time-sensitive** knowledge, LLMs provide correct answers only during specific time periods.
- Limitation of existing group bias analyses:** They average over all time points, thereby overlooking such temporal variability.
- We refer to this type of evaluation as **time-marginalized group bias**.

### ChronoBias Benchmark

- ChronoBias is tailored to **time-conditional group bias** analysis!
  - Group: Geographical regions
  - Date of knowledge update
- ChronoBias consists of QA pairs in the sports domain across 6 geographical regions from 2015 to 2025, each aligned with a gold passage.



### RQ 1. Temporal Group Bias in Vanilla LLMs



### RQ 2. Rationale for Temporal Group Bias in Vanilla LLMs

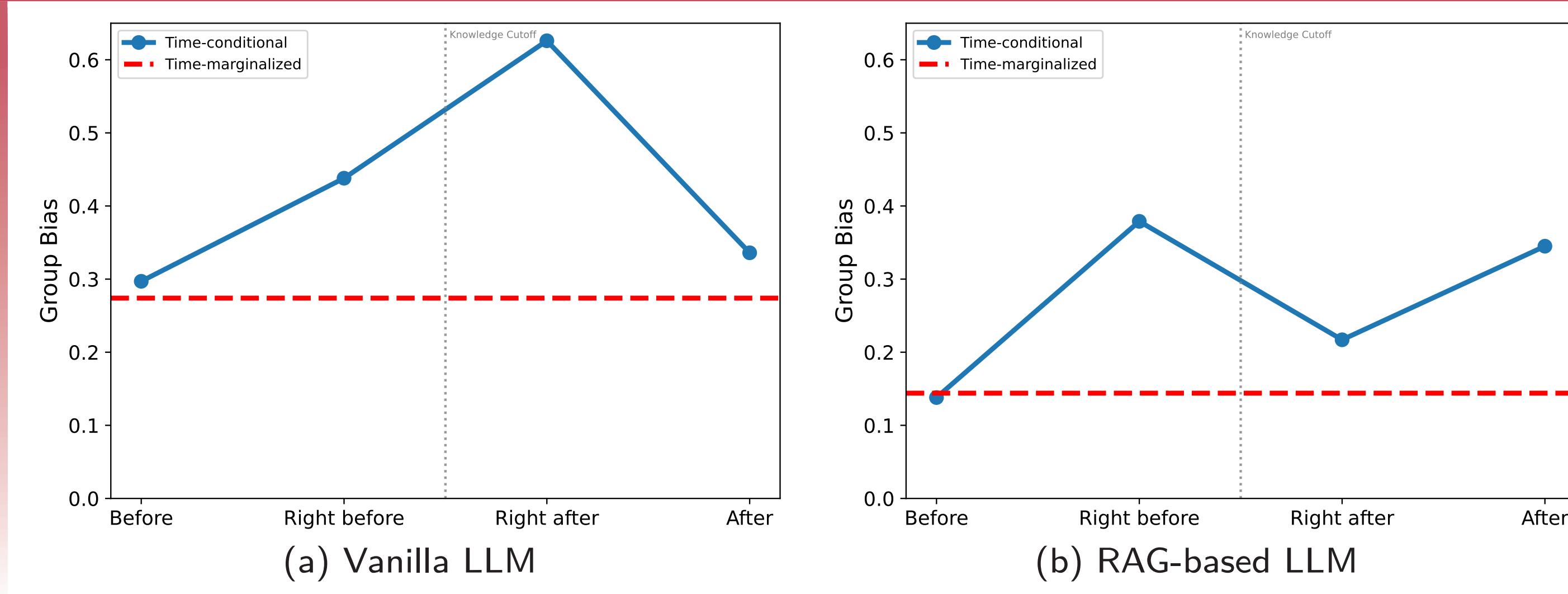
**Parametric knowledge bias:** Before and After the knowledge cutoff

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**Time-sensitivity bias:** After the knowledge cutoff

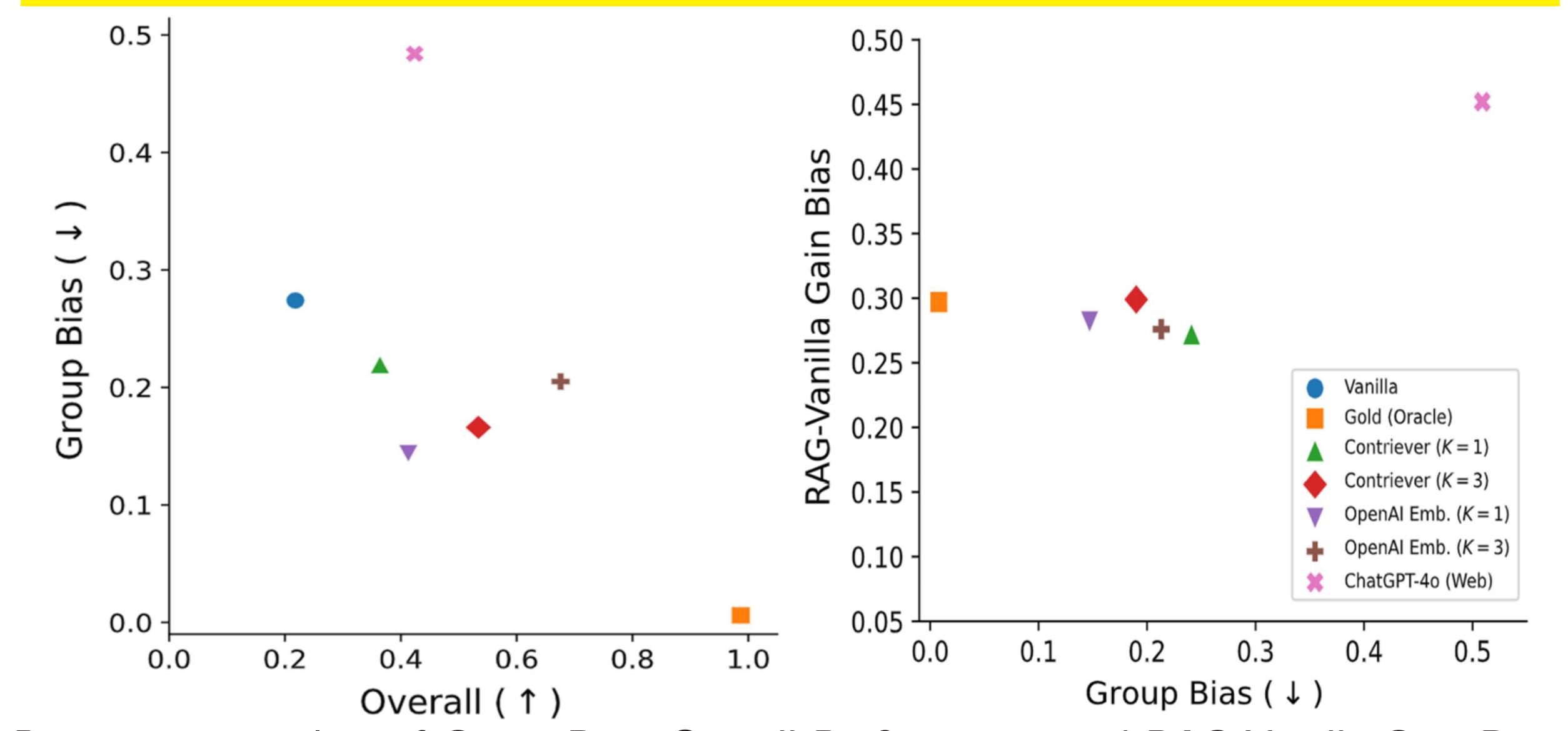
Parametric Knowledge Region	High		Low	
	Slow	Fast	Slow	Fast
Top	1	5.42	4.16	6.42
	3	9.81	9.21	12.76
	5	13.13	15.08	16.84
Bottom	1	9.32	8.63	9.13
	3	21.53	20.37	21.55
	5	27.58	25.05	28.66

### RAG Partially Mitigates Temporal Group Bias



### RQ 3. Guidelines for Building Fair and Accurate RAG-based LLMs

#### Guideline 1: "UNFAIR" retrieval for "FAIR" model performance



**Overall Performance:** Group-wise model performance

**RAG-Vanilla Gain Bias:** Disparity between the maximum and minimum performance improvements from RAG.

**Guideline 2:** Even when the same type of retrieval error occurs, **time-sensitivity bias** causes variation in how much useful partial information is present in the retrieved document.

