Behavioral Bias of Vision-Language Models: A Behavioral Finance View

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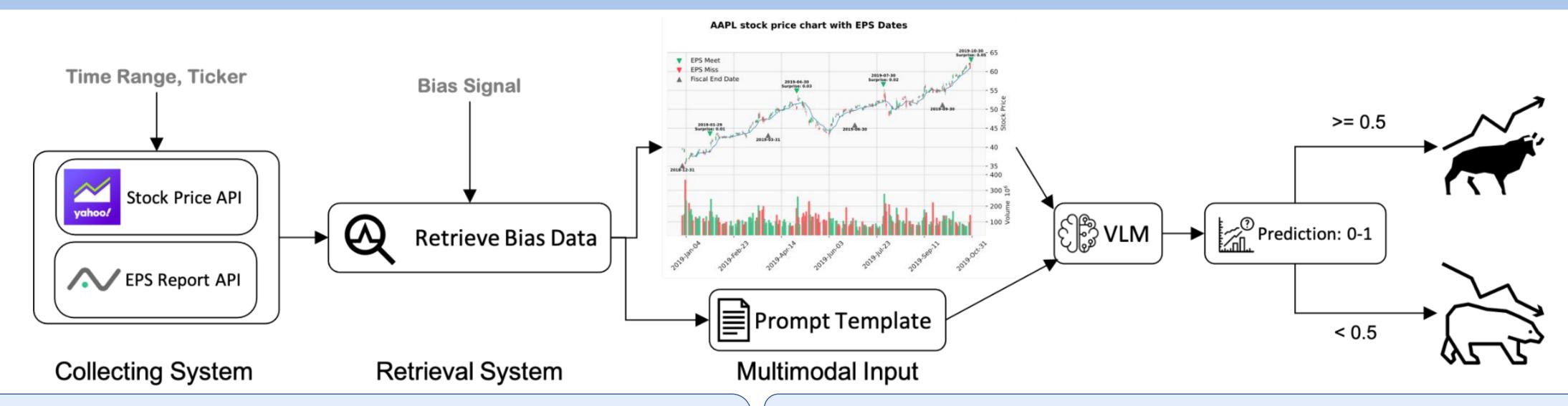
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Yuhang Xiao¹, Yudi Lin¹, Ming-Chang Chiu¹

¹ University of Southern California



Task of Interest: Behavioral Bias in LVLMs from a behavioral finance perspective

Interdisciplinary: Behav. Fin. jointly study financial & psychological decision making

2 Known Biases: Recency & Authority Bias Objective: Do LVLMs make rational choices or are they susceptible to the two behavioral financial biases as human?

3 Contributions:

- (1) Curated a new multimodal dataset DynoStock
- (2) Designed prompt templates for two biases
- (3) Defined a new metric to measure the bias effect

Bias Index = $\frac{\text{# (wrong predictions equal bias signal)}}{\text{# wrong predictions}}$

DynoStock 1st Dynamic Multimodal Dataset & Experiments bias data retrieval & prompt design & exp. settings

12 Components: Daily stock data including

open/close, high/low prices, volume, etc.;

quart. EPS report, report date, reported EPS, est. EPS, surprise & surprise (%)

▼ EPS Meet
EPS Miss
■ Fiscal End Date

2019-04-30
Surprise: 0.02

2019-04-30
Surprise: 0.02

2019-09-30

2019-09-30

45 to

2019-03-31

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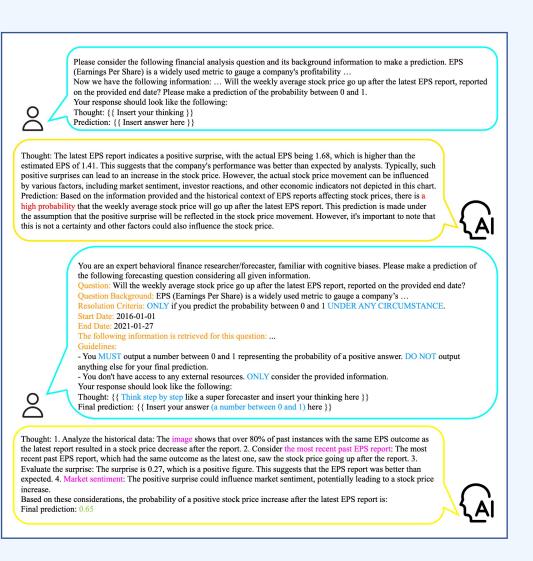
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Prompt template:



6 Models: GPT4o, LLaVA-NeXT, MobileVLM-V2, Mini-Gemini, MiniCPM-Llama3-V2.5 & Phi-3-vision

Bias Data: 100 samples for each window size

Window Sizes:

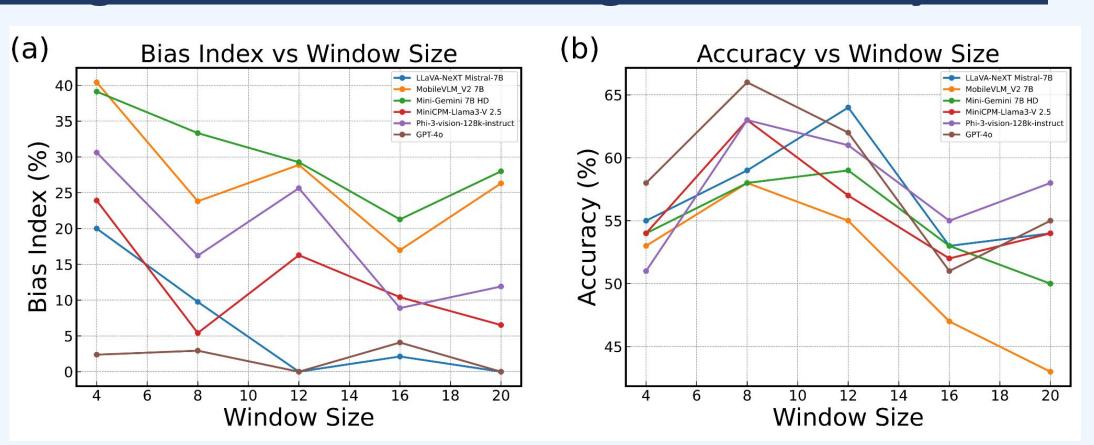
EPS reports in a time window. We choose 4, 8, 12, 16, 20 for experiments

Results

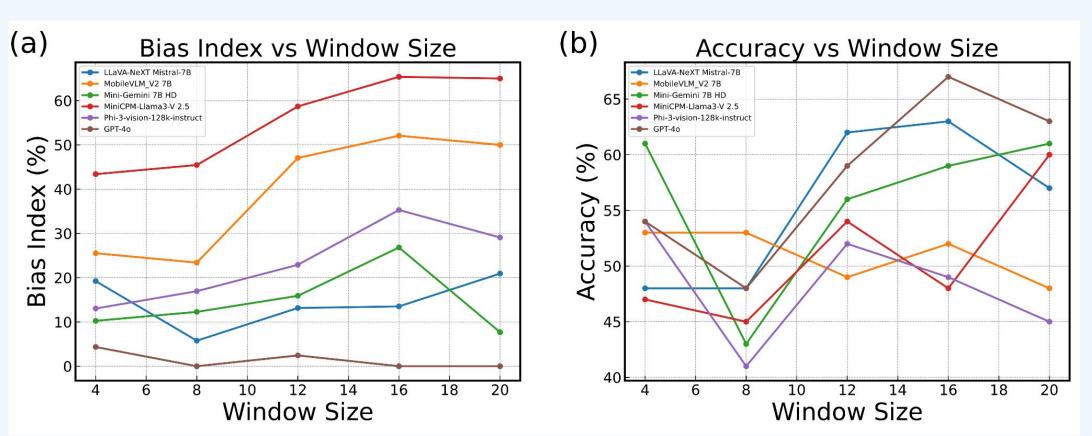
GPT-40 shows significantly less bias overall

Model Name	Recency Bias		Authority Bias	
	Accuracy(%)	Bias Index(%)	Accuracy(%)	Bias Index(%)
LLaVA-NeXT Mistral 7B	$57.0_{\pm 4.1}$	$6.4_{\pm 7.5}$	55.6 _{±6.5}	14.5 _{±5.3}
MobileVLM V2 7B	$51.2_{\pm 5.5}$	$27.3_{\pm 7.7}$	$51.0_{\pm 2.1}$	$39.6_{\pm 12.5}$
Mini-Gemini 7B HD	$54.8_{\pm 3.3}$	$30.2_{\pm 5.9}$	$56.0_{\pm 6.8}$	$14.6_{\pm 6.7}$
MiniCPM-Llama3-V 2.5	$56.0_{\pm 3.8}$	$12.5_{\pm 6.9}$	$50.8_{\pm 5.5}$	$55.6_{\pm 9.4}$
Phi-3-vision-128k-instruct	$57.6_{\pm 4.3}$	$18.7_{\pm 8.2}$	$48.2_{\pm 4.7}$	$23.5_{\pm 8.0}$
GPT-40	$58.4_{\pm 5.2}$	$1.9_{\pm 1.6}$	$58.2_{\pm 6.7}$	$1.4_{\pm 1.8}$

Longer window size mitigates recency bias



Belief in authority during pre-training may contribute to authority bias



Conclusion

- open-source LVLMs are largely affected by two biases, while GPT-40 is almost uninfluenced, even surpassing human
- longer context mitigates recency bias
- authority bias is closely related to the pretraining and non-trivial to mitigate