

Detecting AI Slop in Research and Beyond

Danish Pruthi

Natural Language Processing

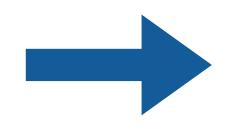
The science and engineering of building computational models to comprehend language

Natural Language Processing

The science and engineering of building computational models to comprehend language

Text Classification

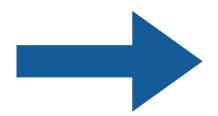
"Lots of epic shows feel a little underpopulated towards the end but there's really no excuse for something as mythic, huge and mesmerizing to end as disappointingly as this."



Negative

Machine Translation

"India recorded their first Test victory, in their 24th match, against England at Madras in 1952. Later in the same year, they won their first Test series, which was against Pakistan."



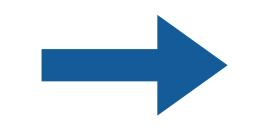
भारत ने 1952 में मद्रास में इंग्लैंड के खिलाफ अपने 24वें मैच में अपनी पहली टेस्ट जीत दर्ज की। बाद में उसी वर्ष, उन्होंने अपनी पहली टेस्ट श्रृंखला जीती, जो पाकिस्तान के खिलाफ थी।

Question answering

"When did India win their first test match?"

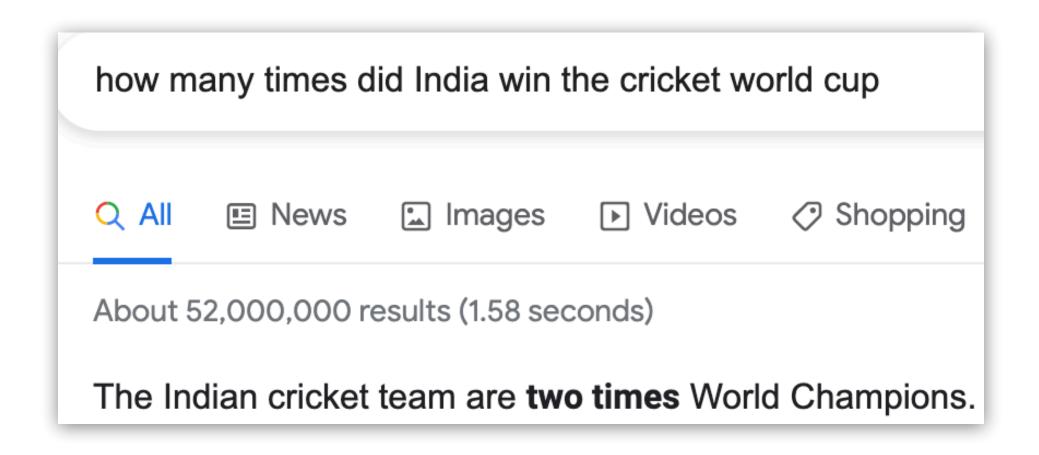




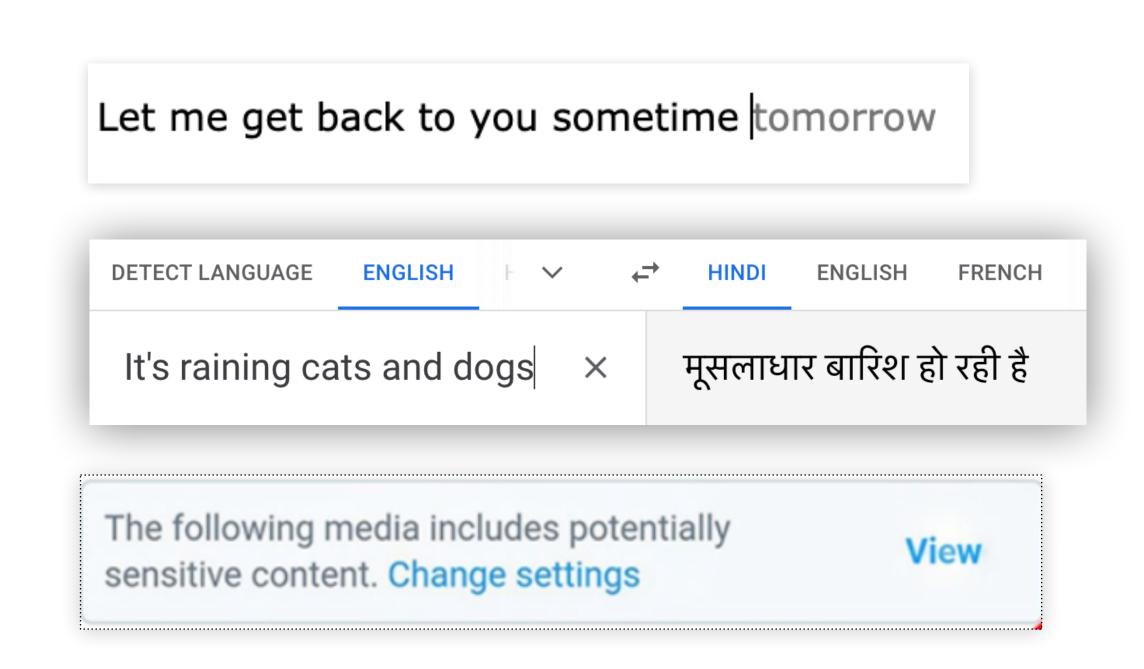


1952

Impact of Language Technologies





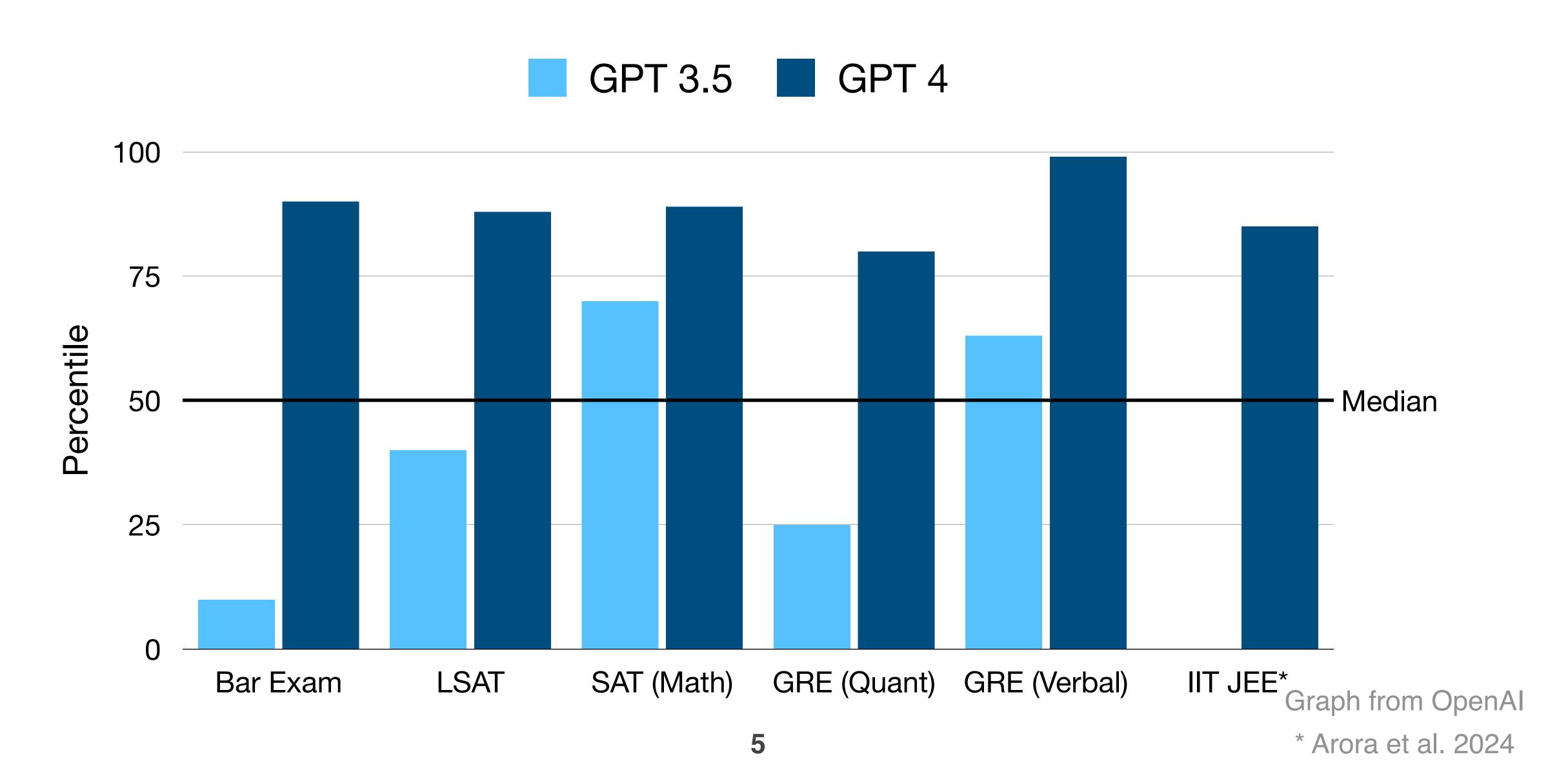


Language Models

Models that assign probabilities to a sequence of words

- I am sorry for the inconvenience ____
- Let me get back to you sometime ____
- I work at IISc, I live in ____
- The Prime Minister of India is ____

Recent Performance Trends



Promise of LLMs

• In advancing science (e.g., Alphafold)

Promise of LLMs

- In advancing science (e.g., Alphafold)
- In automating scientific research?

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- In automating scientific research?

The AI Scientist: Towards Fully Automated Open-Ended Scientific Discovery

Chris Lu^{1,2,*}, Cong Lu^{3,4,*}, Robert Tjarko Lange^{1,*}, Jakob Foerster^{2,†}, Jeff Clune^{3,4,5,†} and David Ha^{1,†}
*Equal Contribution, ¹Sakana AI, ²FLAIR, University of Oxford, ³University of British Columbia, ⁴Vector Institute, ⁵Canada CIFAR AI Chair, [†]Equal Advising

Zochi Publishes A* Paper

#1 Scientific Venue in NLP

Published May 27, 2025

Zochi Achieves Main Conference Acceptance at ACL 2025

Today, we're excited to announce a groundbreaking milestone: Zochi, Intology's Artificial Scientist, has become the first AI system to independently **pass peer review at an A* scientific conference**¹—the highest bar for scientific work in the field.

This achievement marks a **watershed moment** in the evolution of innovation. For the first time, an artificial system has independently produced a scientific discovery and published it at the level of the field's top researchers—making Zochi **the first PhD-level agent**. The peer review process for the main conference proceedings of such venues is designed to be highly selective, with stringent standards for novelty, technical depth, and experimental rigor. To put this achievement in perspective, most PhD students in computer science spend **several years** before publishing at a venue of this stature. AI has crossed a threshold of scientific creativity that allows for contributions alongside these researchers at the highest level of inquiry.

Towards an AI co-scientist

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Juraj Gottweis*, ‡, ¹, Wei-Hung Weng*, ‡, ², Alexander Daryin*,¹, Tao Tu*,³,
Anil Palepu², Petar Sirkovic¹, Artiom Myaskovsky¹, Felix Weissenberger¹,
Keran Rong³, Ryutaro Tanno³, Khaled Saab³, Dan Popovici², Jacob Blum², Fan Zhang²,
Katherine Chou², Avinatan Hassidim², Burak Gokturk¹,
Amin Vahdat¹, Pushmeet Kohli³, Yossi Matias²,
Andrew Carroll², Kavita Kulkarni², Nenad Tomasev³, Yuan Guan²,
Vikram Dhillon⁴, Eeshit Dhaval Vaishnav⁵, Byron Lee⁵,
Tiago R D Costa⁶, José R Penadés⁶, Gary Peltz²,
Yunhan Xu³, Annalisa Pawlosky¹, ‡, Alan Karthikesalingam², ‡ and Vivek Natarajan², ‡

¹Google Cloud AI Research, ²Google Research, ³Google DeepMind,

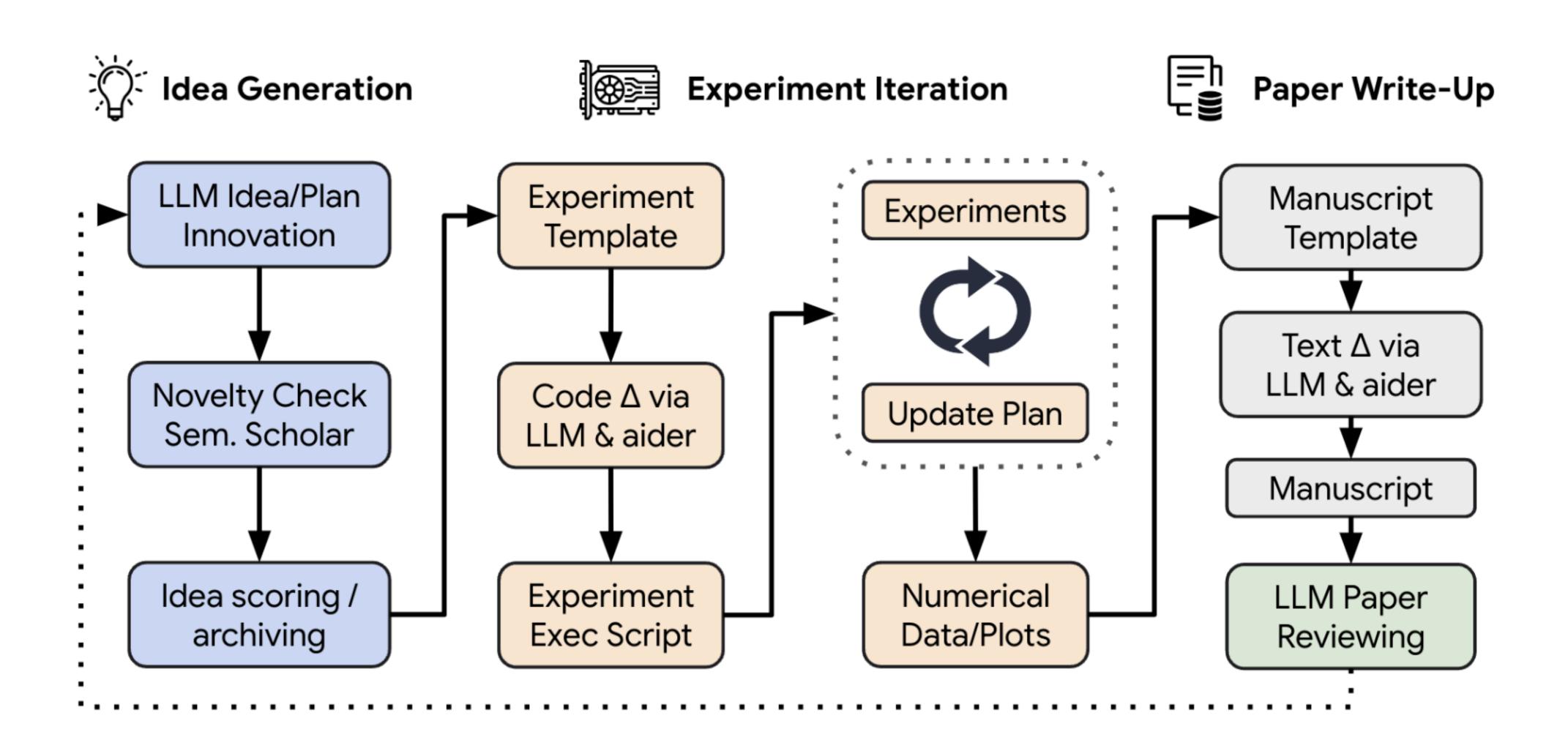
⁴Houston Methodist, ⁵Sequome,

⁶Fleming Initiative and Imperial College London,

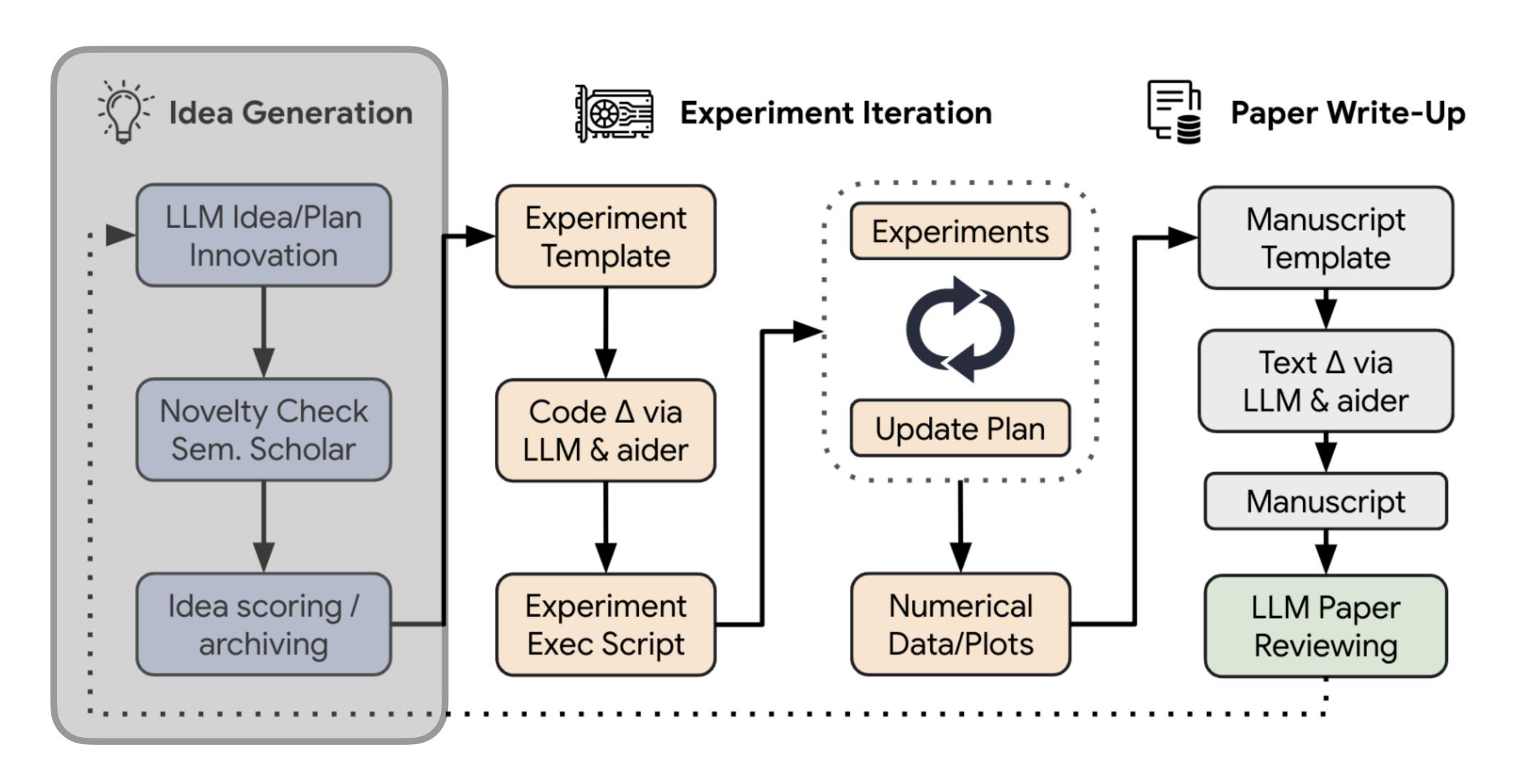
¬Stanford University School of Medicine
```

https://arxiv.org/pdf/2502.18864

Conceptual Illustration of the "AI Scientist"



Conceptual Illustration of the "AI Scientist"



Can LLMs Generate Novel Research Ideas?

A Large-Scale Human Study with 100+ NLP Researchers

Chenglei Si, Diyi Yang, Tatsunori Hashimoto Stanford University

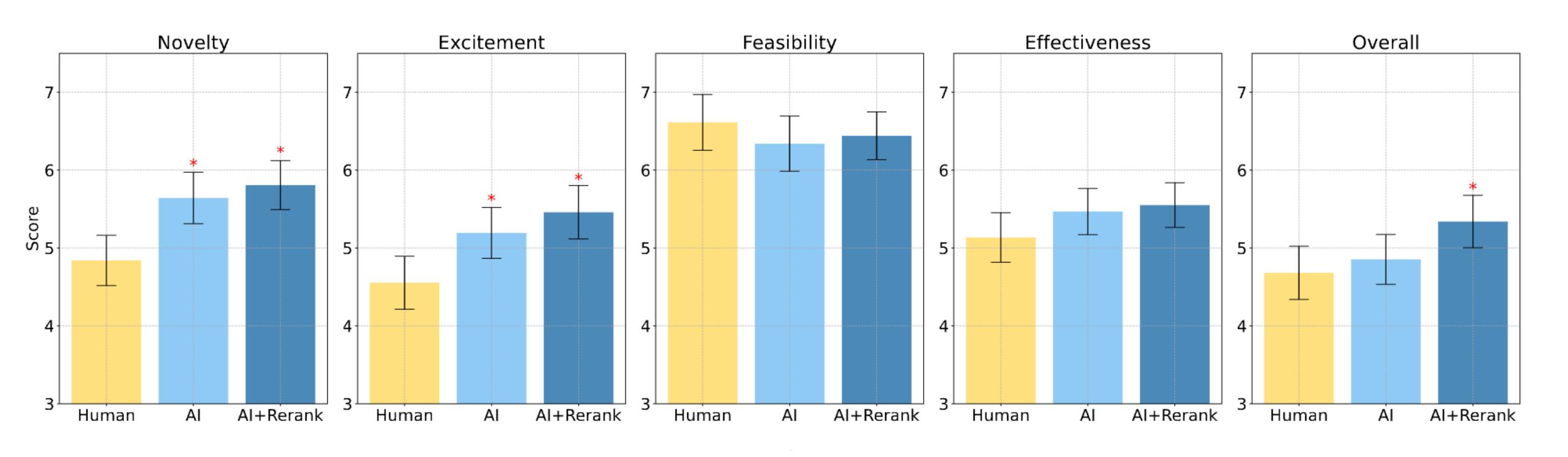
{clsi, diyiy, thashim}@stanford.edu

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Evaluation Philosophy

• Prior work: Experts assess shuffled LLM/human documents for novelty, feasibility, interestingness, etc.

- Our work: Experts actively search for plagiarism
 - Different situational logic (Popper, 2013)
 - Presume plagiarism
 - Actively search for overlap in methodology in existing work

Dataset

• 50 LLM-generated research documents

- 36 fresh proposals generated from Si et al. (2024)
- 4 exemplar proposals from Si et al. (2024)
- 10 exemplar papers from "The AI Scientist" paper (Lu et al. (2024))

• 12 NLP research topics

- Long context capabilities, abstention techniques, bias evaluation
- Hallucination reduction, interpretability, speech processing
- Formal proof generation, human evaluation, machine translation
- Scaling laws, inference optimization, persona development

Expert-led Evaluation

• 13 experts from 5 universities, 2 industrial labs

Score	Description
5	Direct Copy: One-to-one mapping with existing methods
4	Combined Borrowing: Mix-and-match from 2-3 prior works
3	Partial Overlap: Decent similarity, no exact correspondence
2	Minor Similarity: Very slight resemblance, mostly novel
1	Original: Completely novel

Key result

• Large number of proposals (> 24%) are plagiarized

Score	Total Claims (%)	Verified (%)
5	18.0% (9/50)	14.0% (7/50)
4	18.0% (9/50)	10.0% (5/50)
3	32.0% (16/50)	8.0% (4/50)
2	28.0% (14/50)	4.0% (2/50)
1	4.0% (2/50)	0.0% (0/50)

The nature of plagiarism is sophisticated

- Models learn to disguise existing work as novel
- Re-invent terminologies
 - "Resonance graph" instead of "weighted adjacency matrix"

• Several other case studies in the paper

• We create a synthetic dataset of plagiarized ideas

Visual Interpretability for Deep Learning: a Survey

Quanshi Zhang and Song-Chun Zhu University of California, Los Angeles

Abstract

This paper reviews recent studies in understanding neural-network representations and learning neural networks with interpretable/disentangled middlelayer representations. Although deep neural net-works have exhibited superior performance in various tasks, the interpretability is always the Achilles' heel of deep neural networks. At present, deep neural networks obtain high discrimination power at the cost of low interpretability of their black-box representations. We believe that high model interpretability may help people to break several bottlenecks of deep learning, e.g. learning from very few annotations, learning via human-computer communications at the semantic level, and semantically debugging network representa-tions. We focus on convolutional neural networks (CNNs), and we revisit the visualization of CNN representations, methods of diagnosing represen-tations of pre-trained CNNs, approaches for dis-entangling pre-trained CNN representations, learn-ing of CNNs with disentangled representations, and middle-to-end learning based on model inter-pretability. Finally, we discuss prospective trends in explainable artificial intelligence.

1 Introduction

Convolutional neural networks (CNNs) [LeCun et al., 1998a; Krizhevsky et al., 2012; He et al., 2016; Huang et al., 2017]
have achieved superior performance in many visual tasks,
tions is the most direct way to explore network representasuch as object classification and detection. However, the endto-end learning strategy makes CNN representations a black box. Except for the final network output, it is difficult for people to understand the logic of CNN predictions hidden inside to-end learning strategy makes CNN representations a black foundation for many approaches to diagnosing CNN representations. The disentanglement of feature representations of a pre-trained CNN and the learning of explainable network the network. In recent years, a growing number of researchers have realized that high model interpretability is of significant value in both theory and practice and have developed models resentations are also the starting point for weakly-supervised

with interpretable knowledge representations.

In this paper, we conduct a survey of current studies in unral networks with interpretable/disentangled representations.

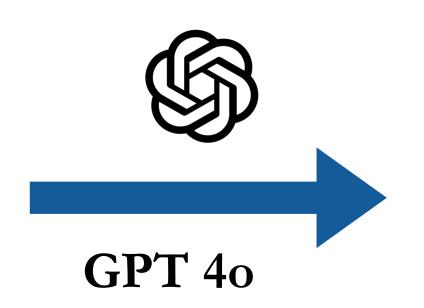
We can roughly define the scope of the review into the fol-

- Visualization of CNN representations in intermediate network layers. These methods mainly synthesize the image that maximizes the score of a given unit in a pretrained CNN or invert feature maps of a conv-layer back to the input image. Please see Section 2 for detailed dis-
- Diagnosis of CNN representations. Related studies may either diagnose a CNN's feature space for different object categories or discover potential representation flaws in conv-layers. Please see Section 3 for details. · Disentanglement of "the mixture of patterns" encoded
- tangle complex representations in conv-layers and transform network representations into interpretable graphs. · Building explainable models. We discuss interpretable
- CNNs [Zhang et al., 2017c], capsule networks [Sabour et al., 2017], interpretable R-CNNs [Wu et al., 2017], and the InfoGAN [Chen et al., 2016] in Section 5.
- Semantic-level middle-to-end learning via humancomputer interaction. A clear semantic disentanglement of CNN representations may further enable "middle-toend" learning of neural networks with weak supervision. Section 7 introduces methods to learn new models via human-computer interactions [Zhang et al., 2017b] and active question-answering with very limited human supervision [Zhang et al., 2017a].

tions. The network visualization also provides a technical representations present bigger challenges to state-of-the-art algorithms. Finally, explainable or disentangled network rep-

middle-to-end learning.

Values of model interpretability: The clear semantics in derstanding neural-network representations and learning neu-As discussed in [Zhang et al., 2018b], considering dataset and representation bias, a high accuracy on testing images still cannot ensure that a CNN will encode correct represen





Plagiarized Research Article

Method		Accuracy
	Oracle access	88.8%
Claude 3.5 Sonnet	Parameteric Knowledge	1.3%
	SSAG	51.3%
	Oracle access	89.0%
GPT-4o	Parameteric Knowledge	32.7%
	SSAG	68.5%
OpenScholar		0%
Turnitin		0%

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• Experts only looked through AI-generated research;

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- But experts regularly peer-review papers...
 - So we extract signs of plagiarism in peer-reviews

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Conference	Score 4	Score 5	Plagiarism rate (scores 4+)
	(%)	(%)	(%)
ACL 2017	0.8%	0%	0.8%
ICLR 2017	4.0%	2.3%	6.3%
CoNLL 2016	5.3%	0%	5.3%
NeurIPS 2017	1.8%	0%	1.8%

Broader Implications of AI Scientists

- Regurgitate old ideas, without any attribution
- Overwhelm conferences
- Fracture scientific discourse
- Sow distrust about other legitimate AI capabilities

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All That Glitters is Not Novel: Plagiarism in AI Generated Research

By Tarun Gupta, Danish Pruthi

ACL 2025



Impact of the work

• Paper received the outstanding paper award at ACL

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• Paper received the outstanding paper award at ACL



What counts as plagiarism? Algenerated papers pose new risks

Researchers argue over whether 'novel' AI-generated works use others' ideas without credit.

By <u>Ananya</u>

Broadly: Detecting AI Slop

AI Slop: low-quality, often nonsensical or misleading, content generated by artificial intelligence.

With 'AI slop' distorting our reality, the world is sleepwalking into disaster

By Nesrine Malik in the Guardian

AI-generated 'slop' is slowly killing the internet

By Arwa Mahdawi in the Guardian

Increasing Reports of Plagiarism

Business News / Ai / Artificial Intelligence / UK universities launch probe after 400 stud...

UK universities launch probe after 400 students found cheating through ChatGPT

1 min read • 07 Jul 2023, 07:34 PM IST

Edited By **Devesh Kumar**

My students are using AI to cheat. Here's why it's a teachable moment Siva Vaidhyanathan EXCLUSIVE: 'Half of school and college students are already using ChatGPT to cheat': Experts warn Al tech should strike fear in all academics

- Many school districts have already banned the use of ChatGPT
- GPT-4 can score 90 percent on many exams already including the American bar

Increasing Concerns of Targeted Misinformation

Technology

OpenAl chief concerned about Al being used to compromise elections

By Diane Bartz, Zeba Siddiqui and Jeffrey Dastin

May 17, 2023 3:42 AM GMT+5:30 · Updated 5 months ago



AI-generated disinformation poses threat of misleading voters in 2024 election

Politics May 14, 2023 7:52 PM EDT

Need to Distinguish LLM Outputs from Human Text

- Detect plagiarism
- Combat targeted large-scale misinformation/spam/abuse
- Avoid re-training on LLM outputs

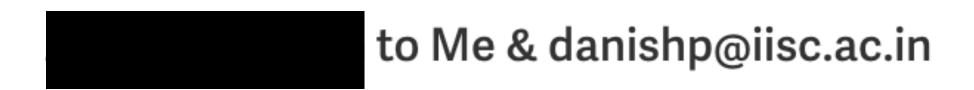
• You might get lucky:

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I am impressed by the innovative work being carried out at the LOCA lab and am particularly drawn to [mention a specific project or aspect of the lab's work that interests you]. My enthusiasm for this research area, combined with my [mention any additional qualifications or achievements], motivates me to contribute to and excel in your research team.

I have attached my CV and any other required documents as per the IISc application guidelines. I am available for interviews at your convenience and am excited about the opportunity to discuss how my skills and experiences align with the goals of the LOCA lab.

• You might get lucky:





It looks like you have applied for a position (possibly a research internship or winter school) at IISc and received a response from Danish. Since you've already applied, you might want to send a polite follow-up email to express your enthusiasm and confirm your application status. Here's a refined response you could use:

Respected sir,

I have officially submitted my application through the IISc admissions portal and also filled out the additional form.

• Model developers could store all the responses generated

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• Model developers could store all the responses generated

- Wouldn't work for open-source models
- Privacy concerns

Possible Approaches: Watermarking



A Watermark for Large Language Models

John Kirchenbauer * Jonas Geiping * Yuxin Wen Jonathan Katz Ian Miers Tom Goldstein
University of Maryland

ICML 2023 (Best Paper Award)

I work at the Indian Institute of Science. I live in _____

```
I work at the Indian Institute of
   Science. I live in Bangalore 0.7
                          the 0.1
                        Chicago 0.05
                        Seattle 0.04
                       California 0.005
                         India 0.004
                        London 0.003
                        Canada 0.001
```

```
I work at the Indian Institute of
   Science. I live in
                       Bangalore
                                 0.7
                                0.1
                          the
                        Chicago 0.05
                        Seattle 0.04
                       California 0.005
                         India 0.004
                         London 0.003
                                0.001
                         Canada
```

- 1. Randomly partition the vocabulary based on the last word
- 2. Don't generate a word from the red list

I work at the Indian Institute of
Science. I live in Bangalore 0.7
the 0.1
Chicago 0.05
Seattle 0.04
California 0.005
India 0.004
London 0.003
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- 1. Randomly partition the vocabulary based on the last word
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I work at the Indian Institute of Science. I live in Bangalore 0.1 the Chicago 0.05 Seattle 0.04 California 0.005 India 0.004 London 0.003 0.001 Canada

- 1. Randomly partition the vocabulary based on the last word
- 2. Don't generate a word from the red list

```
I work at the Indian Institute of
   Science. I live in Bangalore
                                            + delta
                                    0.1
                             the
                                            + delta
                           Chicago 0.05
                                            + delta
                           Seattle 0.04
                                            + delta
                          California 0.005
                            India
                                    0.004
                                            + delta
                            London 0.003
                                            + delta
                                    0.001
                            Canada
                                            \bullet \bullet \bullet
```

- 1. Randomly partition the vocabulary based on the last word
- 2 Don't generate a word from the red list Boost green words

Soft

```
I work at the Indian Institute of
   Science. I live in Bangalore
                                             + delta
                                     0.1
                              the
                                             + delta
                            Chicago 0.05
                                             + delta
                            Seattle 0.04
                                             + delta
                          California
                                     0.005
                             India
                                     0.004
                                             + delta
                                    0.003
                            London
                                             + delta
                                     0.001
                            Canada
                                             \bullet \bullet \bullet
```

- 1. Randomly partition the vocabulary based on the last word
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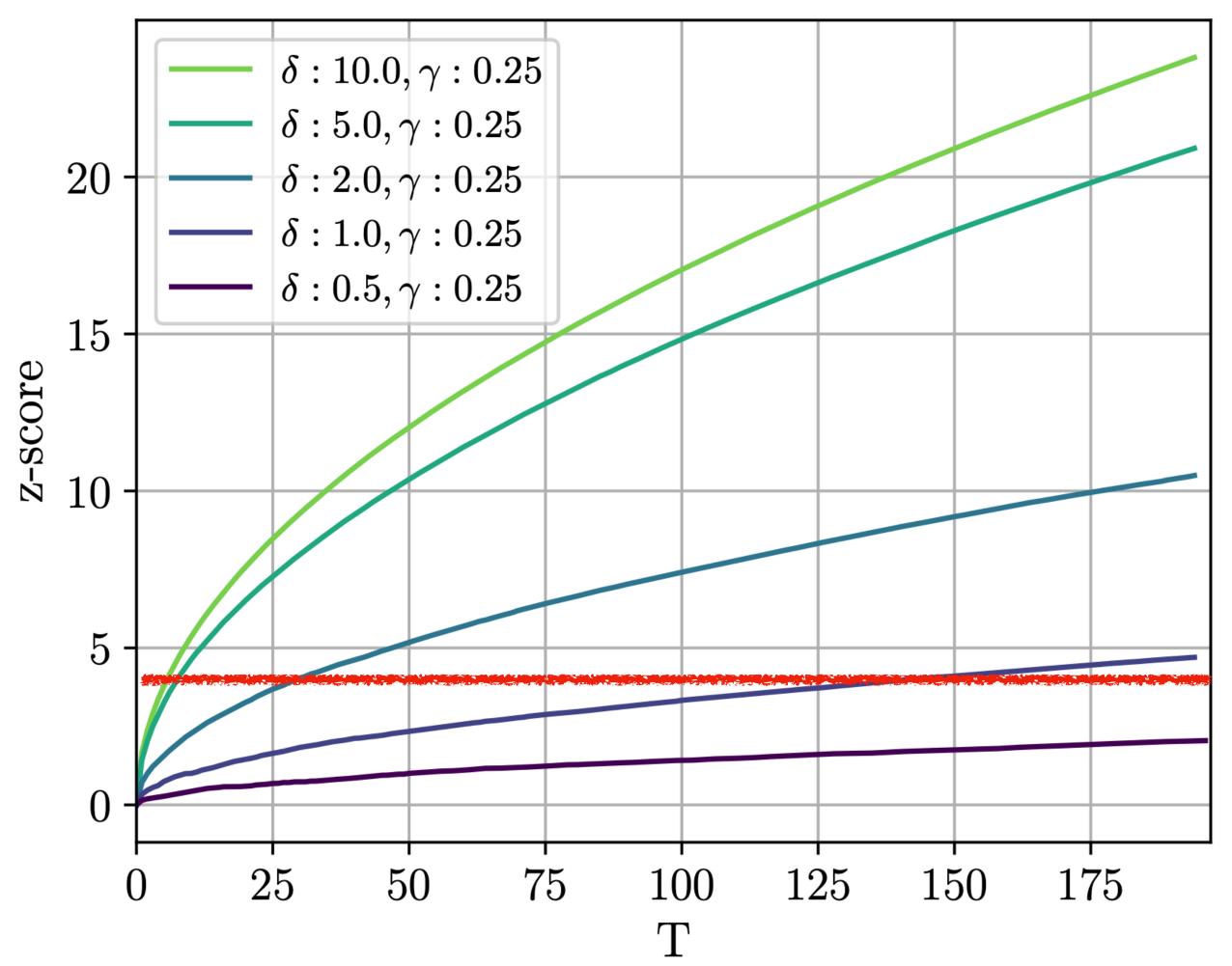
```
large fraction
(than what we would expect randomly)
```

```
large fraction
(than what we would expect randomly)
suspect
(we can run a statistical test)
```

Useful Properties

- Can be applied to any language model
- Knobs to play around with the watermarking strength
- Watermarking is conceptually simple & computationally cheap
- Detection does not depend on the model probabilities

Efficacy of watermarking

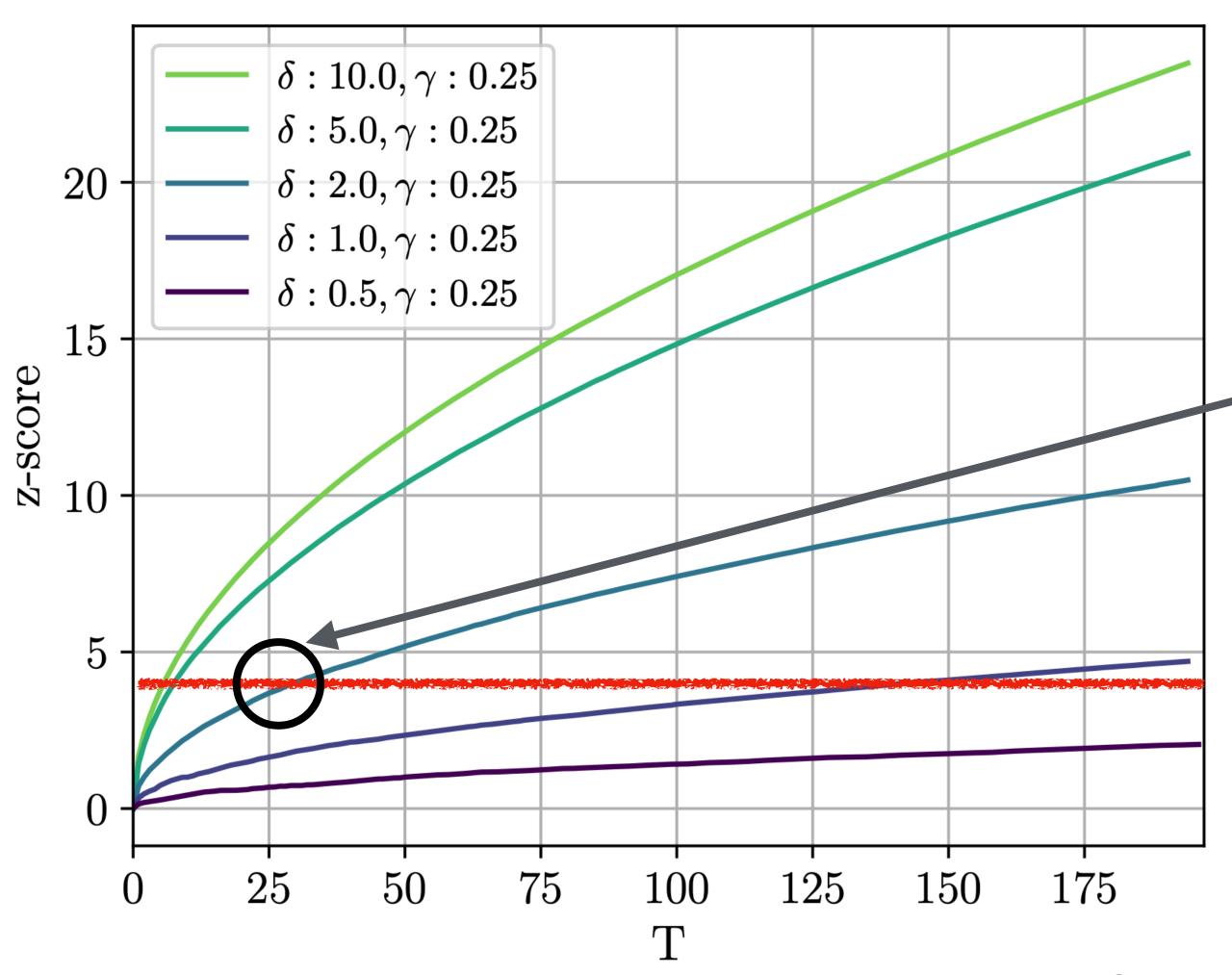


$$z = (|x_G| - \gamma T)/\sqrt{T\gamma(1 - \gamma)}$$

$$z \propto r_G \sqrt{T}$$

False positive rate is 3×10^{-5}

Efficacy of watermarking



$$z = (|x_G| - \gamma T)/\sqrt{T\gamma(1 - \gamma)}$$

$$z \propto r_G \sqrt{T}$$

Often, about 30 tokens suffice

False positive rate is 3×10^{-5}

From Kirchenbauer et al.

Our Work

- Downstream effects of watermarking (Findings of EMNLP, 2024)
 - By Anirudh Ajith, Sameer Singh, Danish Pruthi
 - https://arxiv.org/abs/2311.09816
- Undoing (or reverse-engineering) watermarking (EMNLP 2024)
 - By Saksham Rastogi, Danish Pruthi
 - https://arxiv.org/abs/2411.05277
- Watermarking your own content (ICML 2025)
 - By Saksham Rastogi, Pratyush Maini, Danish Pruthi
 - https://arxiv.org/abs/2504.13416

In Practice: Almost No Adoption

In Practice: Almost No Adoption

• Corporations are (allegedly) worried that they'll lose customers

• Requires all developers to watermark

In Practice

- Hundreds of available detectors:
 - Originality,
 - GPTZero,
 - DetectGPT,
 - Pangram, etc.

• Which work with varying effectiveness

Thank you



Danish Pruthi

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Papers: http://bit.ly/danish037

Email: <u>danishp@iisc.ac.in</u>