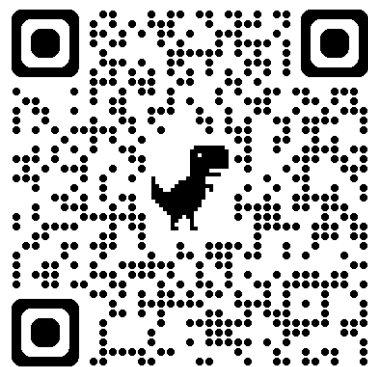


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# Watermarking for Large Language Models

## Part I: Introduction



Xuandong Zhao

UC Berkeley



Yu-Xiang Wang

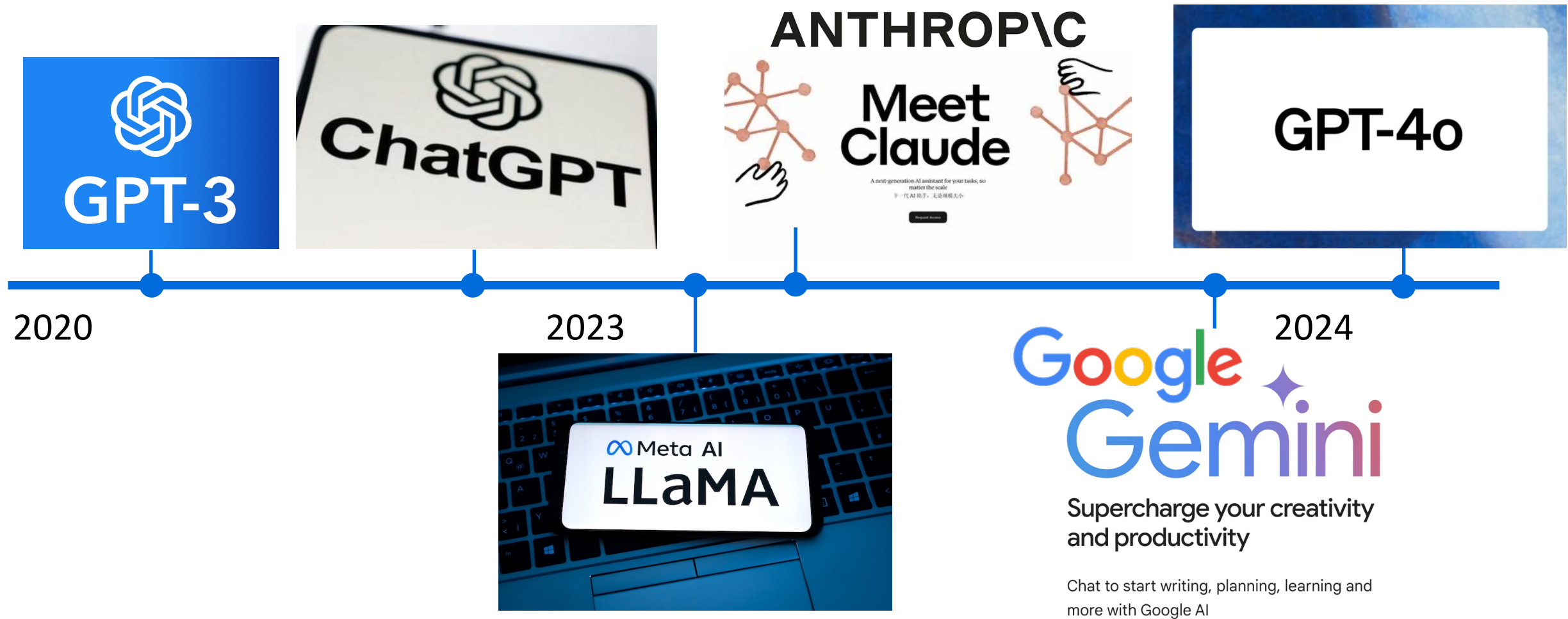
UC San Diego

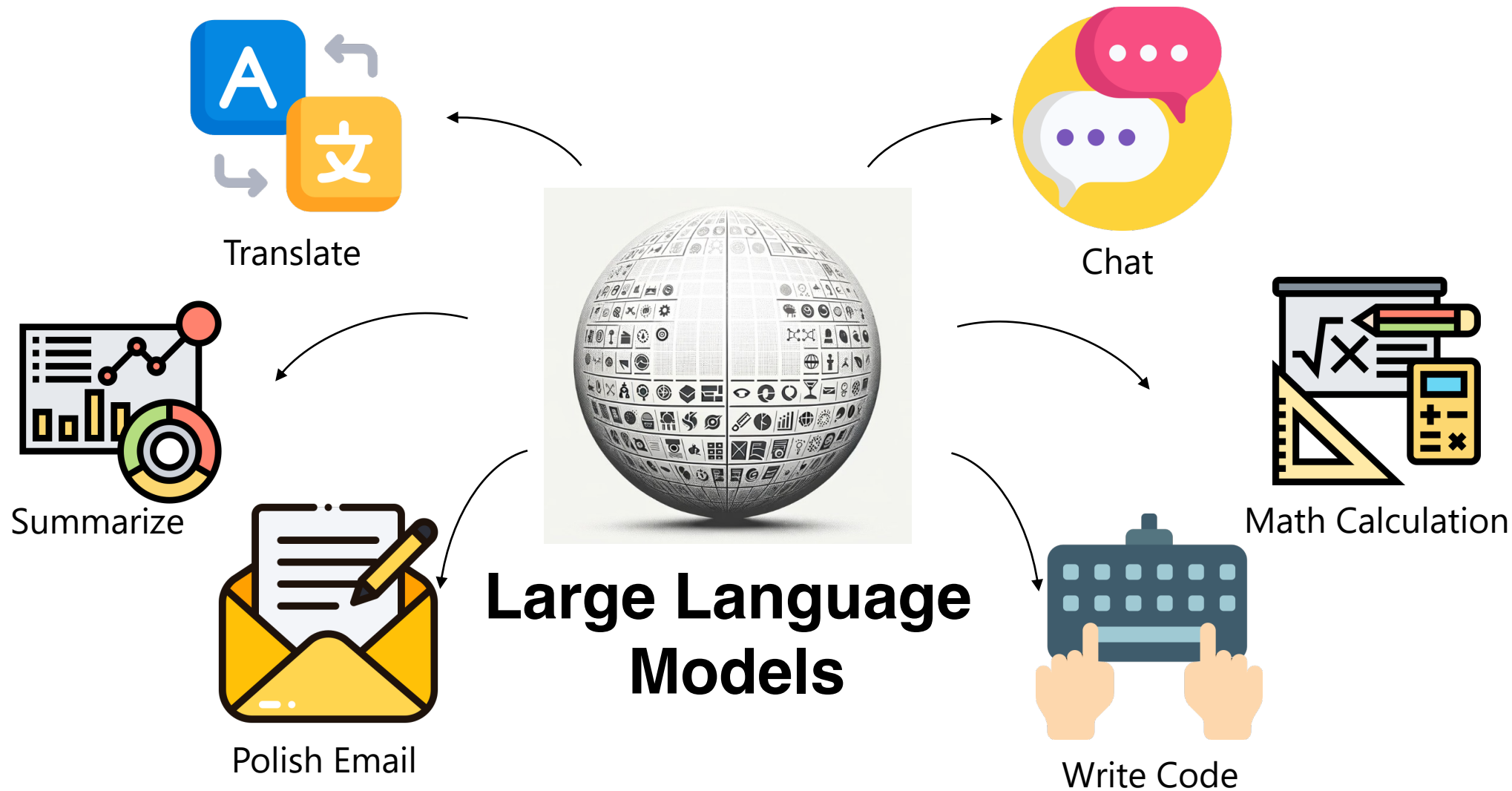


Lei Li

CMU

# Large Language Models





# How people use LLMs/Chatbots?

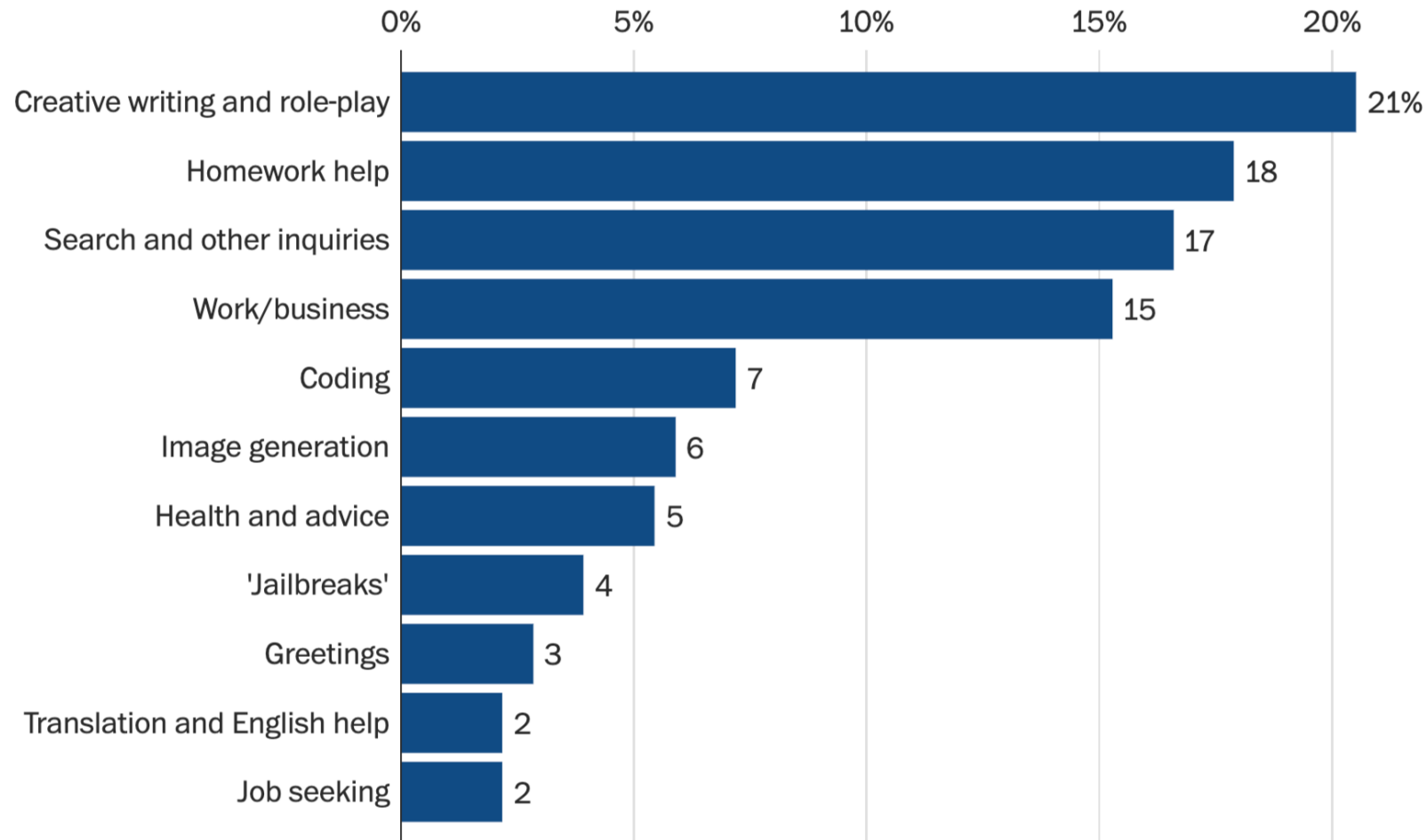


Chart shows proportion of prompts in the category from a random sample of 458 English WildChat conversations, selected from the first prompt per day per US-based IP address. Margin of sampling error is 5 percentage points.

# Risks of LLMs

- Fake news...
- Bogus case law...
- Malware...
- Scams...
- Plagiarism...
- Private data leaks...
- ...

China reports first arrest over fake news

BREAKING  
Judge Fines Two Lawyers For Using

Artificial Intelligence

**ChatGPT Leaks Some User Data, OpenAI Suspects**

*The leaks exposed conversations, pe*

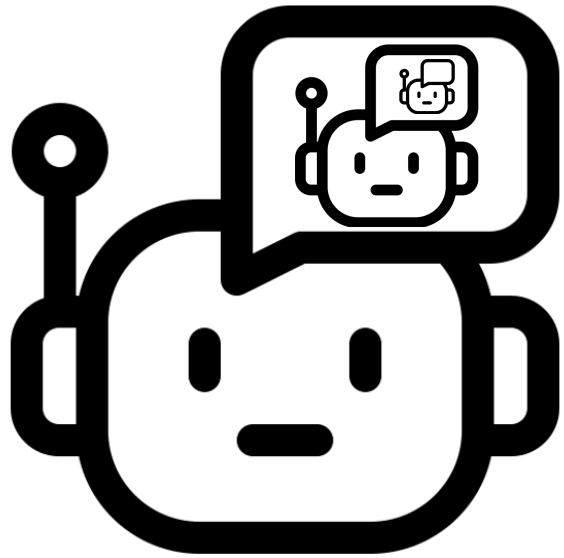
Chris Westfall Contributor @  
*Guidance for leaders and aspiring leaders, interested in career impact*

**Forbes** Follow

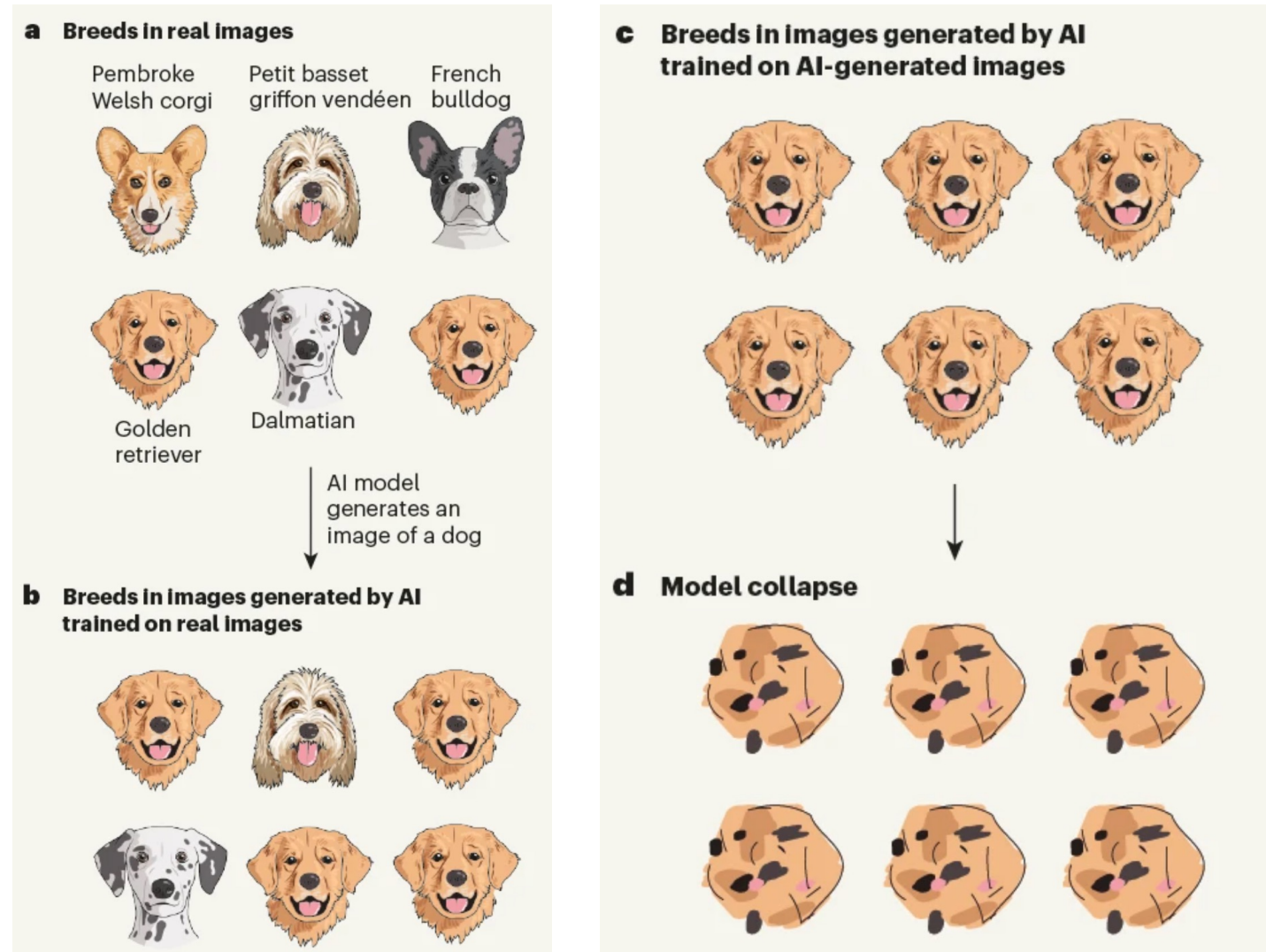
# Why do we need to detect AI-generated text?



# Why do we need to detect AI-generated text?



## Model Degeneration or Model Collapse





# Why do we need to detect AI-generated text?



Executive (and Trust)  
Use of

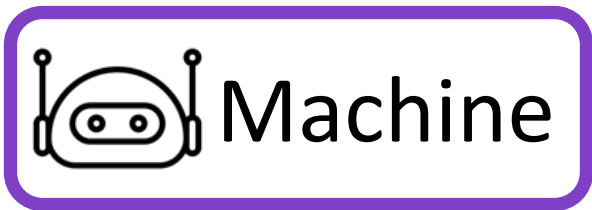
means us  
managin


A screenshot of a social media post from Google DeepMind. The post features a collage of images: a butterfly, a jellyfish, a sunset, and a landscape with a river. The text of the post includes the title "SynthID" and the subtitle "Identifying AI-generated content with SynthID". There is a "Share" button and a play button icon at the bottom right. The background of the post is a blurred image of a document with some text visible, including phrases like "credible speakers who will be sh...", "leaders in their field and hav...", "we will also have other enga...", "sessions and networking op...", "opportunity to dive deeper into...", "relationships.", "TECHNOLOGY", "great success, and I'd love to", "that means", "n or has", "note bias and", "Finally, it", "ange, and".



# Can you distinguish human vs. machine generated text?

Through the town, and past the lights,  
Oh, how the bells do ring!  
They chime with glee  
For you and me  
As carols we joyfully sing.



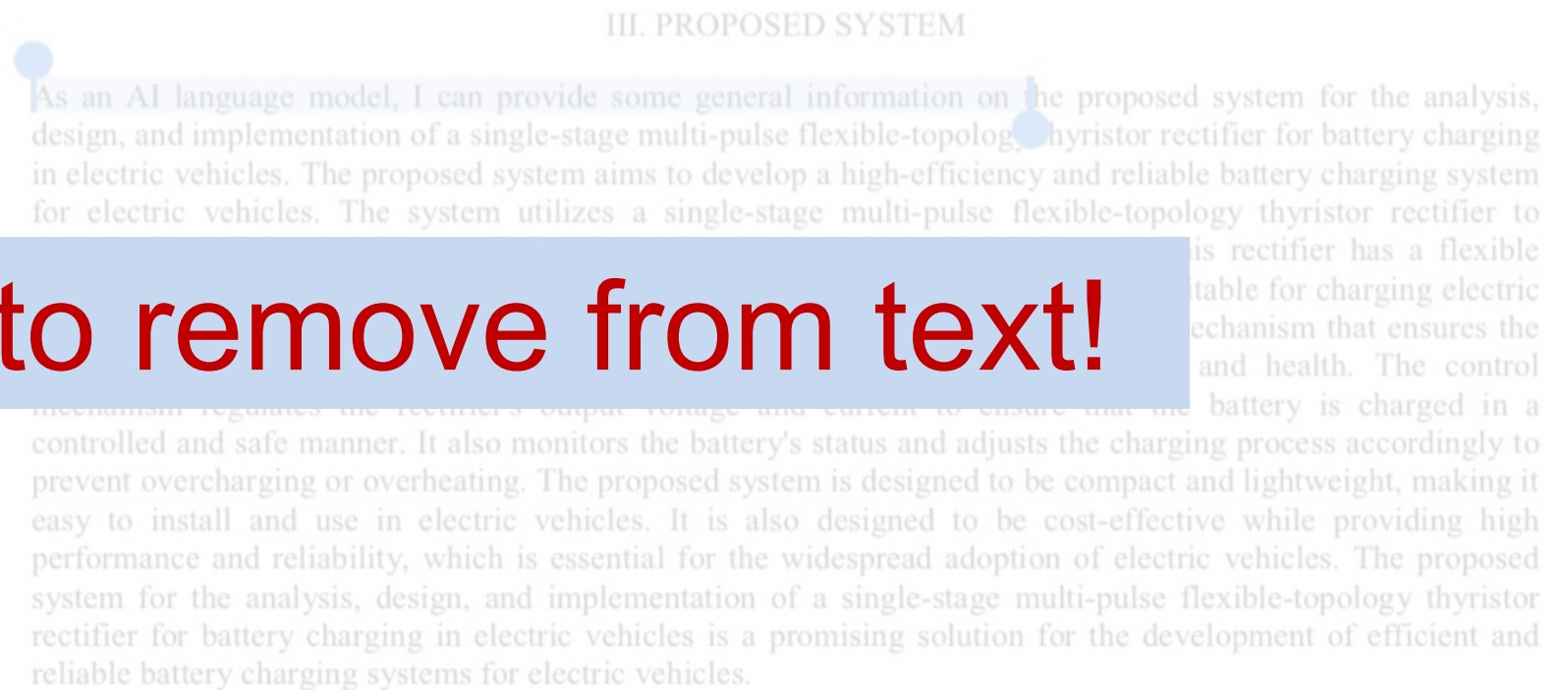
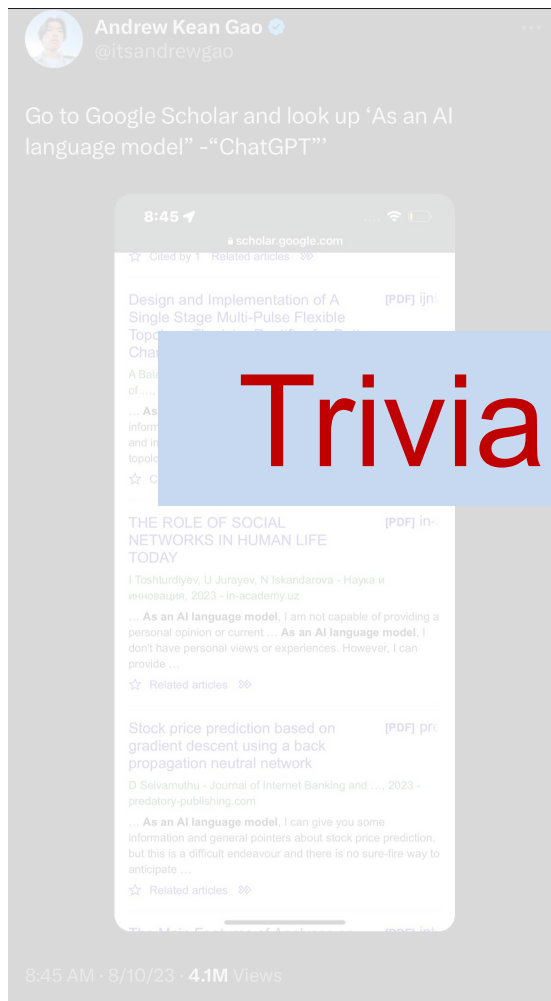
Over the river, and through the wood,  
Oh, how the wind does blow!  
It stings the toes  
And bites the nose  
As over the ground we go. 



Child, Lydia Maria. "Thanksgiving Day." 1844.

# How to detect AI-generated text?

- Add prefix: “As a large language model...”



# How to detect AI-generated text?

- Maintain a database of all completions

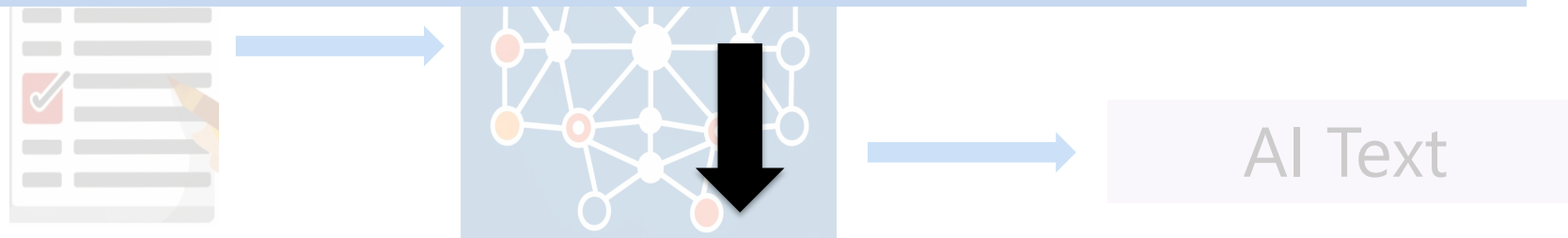
Database of completions



# How to detect AI-generated text?

- Train classification models [GPTZero, Turnitin, ...]

Too many false positives?  
Out-of-distribution data?



Part IV: Post-Hoc Detection

# Watermarking is a promising solution!

Plant subtle but distinctive signals deliberately within the content to enable downstream detection

## Part II: Text Watermarking

Watermarking vs. AI Classifier

Active

Passive

# Intellectual Property of LLM

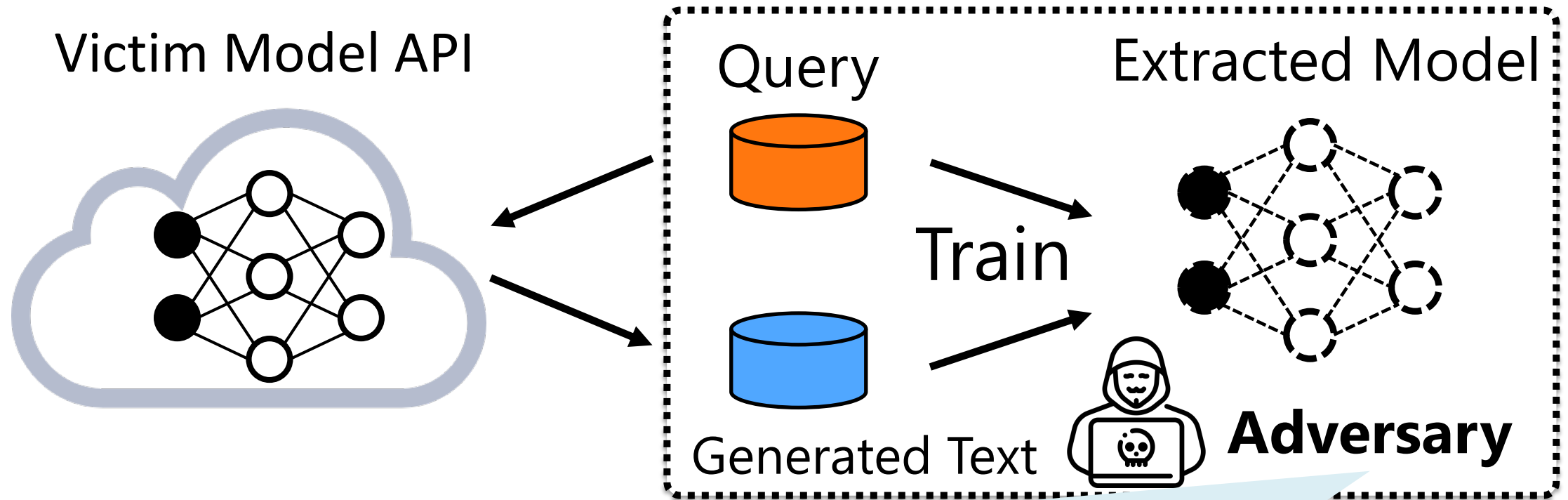


I want to steal the model!



# Model Stealing/Extraction Attack

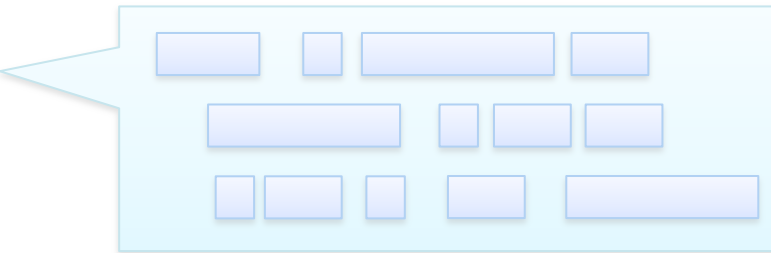
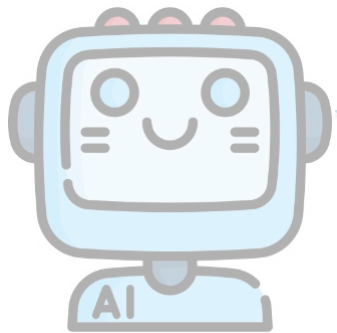
Extract the model information by querying the model in a black-box setting



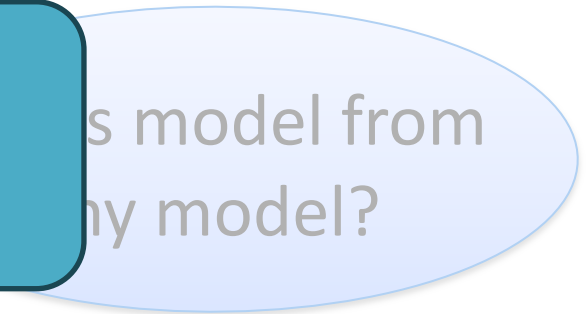
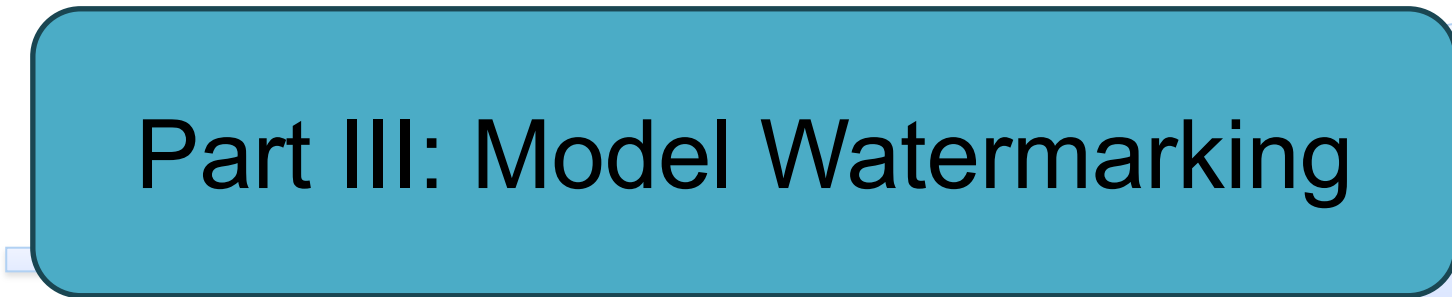
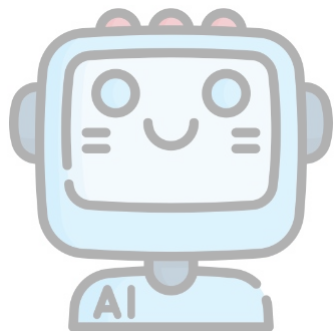
I can obtain a similar model to yours at a much lower cost

# Can we watermark the model?

## Text Watermark



## Model Watermark



# Outline

- Part I: Introduction
- Part II: Text Watermark
  - (a) Green-Red Watermark
  - (b) Gumbel Watermark
  - (c) Theoretical results

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- Part III: Model Watermark
- Part IV: Post-Hoc Text Detection
- Part V: Conclusion and Future Directions

Break

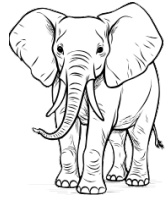


@ Xuandong Zhao

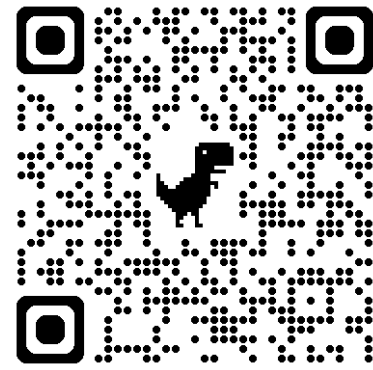


@ Yu-Xiang Wang

@ Lei Li



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# Watermarking for Large Language Models

## Part II: Text Watermarking



Xuandong Zhao  
UC Berkeley



Yu-Xiang Wang  
UC San Diego



Lei Li  
CMU

# Watermarking has a long history



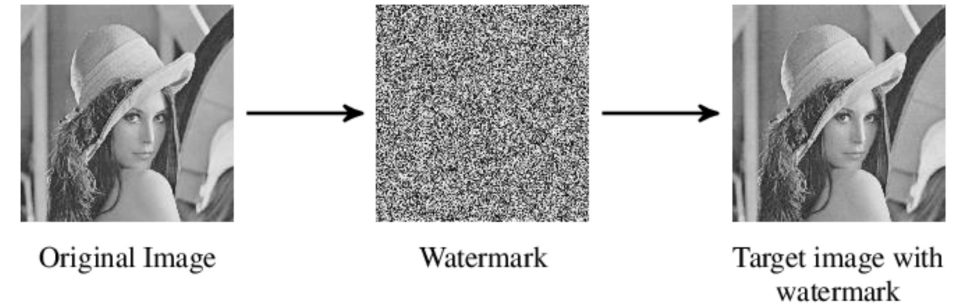
The *Crown CA* watermark found on many British Commonwealth stamps

<https://en.wikipedia.org/wiki/Watermark>

## Traditional Image Watermarks



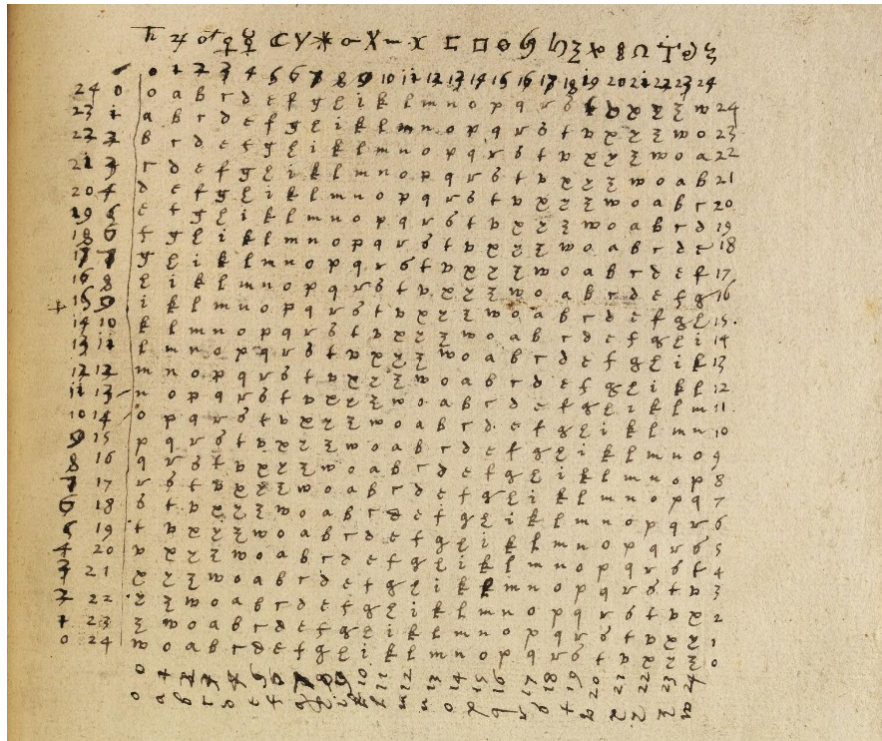
## Invisible Image Watermarks





# Text Watermarking

- Ancient Greece: Steganography
- 1499: Trithemius “Steganographia”



<https://en.wikipedia.org/wiki/Steganography>

- 1950s: Embedding code to music (Hembrooke, 1954)
- 1990s to 2000s: Digital Watermarks (e.g., Ingemar J. Cox, Matt Miller, etc..)
- Rule-based parsed syntactic tree (Atallah et al., 2001)
- Rule-based semantic structure of text (Atallah et al., 2000; Topkara et al., 2006)
- Neural steganography with DL models (Fang et al., 2017; Ziegler et al., 2019)



# 2022+: Recent Renaissance due to the rise of Generative AI

- Watermarking LLM text

Aaronson (2022), Kirchenbauer et al. (2023), Zhao et al. (2023; 2024), Christ et al. (2023), Kuditipudi et al. (2023), Hu et al. (2023), Christ and Gun (2024)

**Part 2 of the tutorial**

- Watermarking LLM models

Zhao et al. (2022) “Distillation resistant watermarking”, Zhao et al. (2023) “Protecting Language Generation Models via Invisible Watermarking”

- Watermarking Images (e.g. from Diffusion models)

(e.g., Fernandez et al. 2023 “Stable signature”, Wen et al. 2023 “Tree-Ring Watermarks”)

- “Is strong watermarking possible?”

“Zhao et al. (2023) Invisible Image Watermarks Are Provably Removable Using Generative AI

Zhang, Barak et al. (2024) Watermarks in the Sand: Impossibility of Strong Watermarking for Generative Models

Sadasivan et al. (2023) Can AI-generated text be reliably detected?

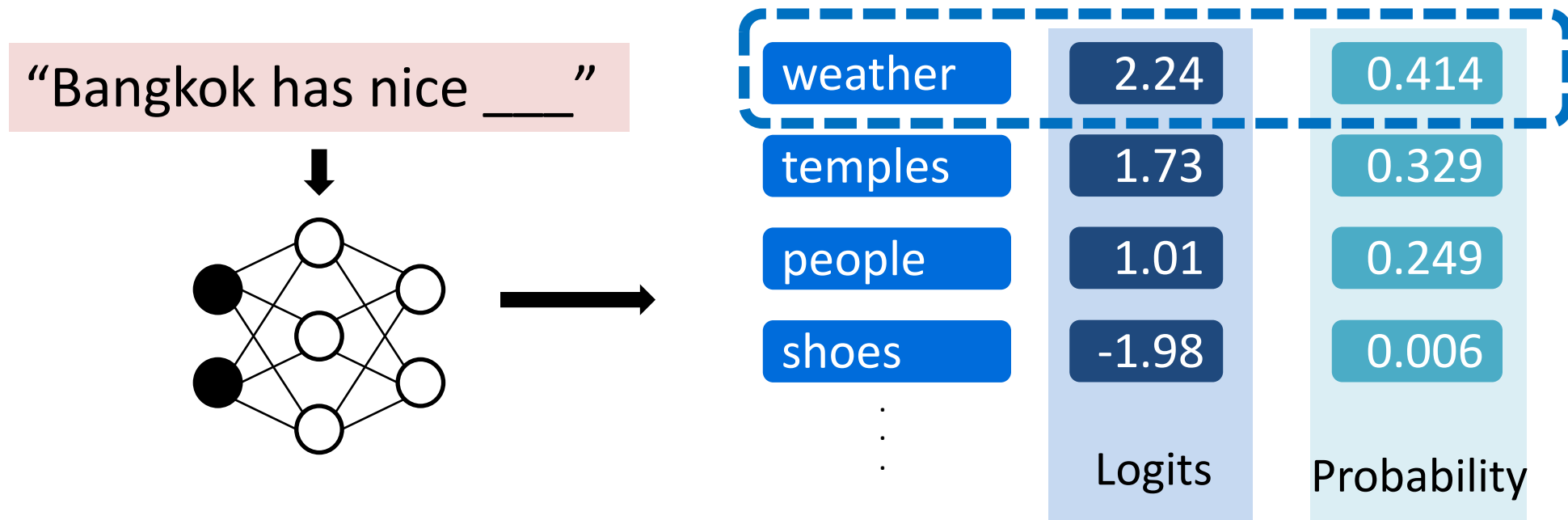
**Slightly different settings, motivating applications and new challenges.**

# Main difference

- Steganography / Watermarking in the 1990s to 2000s
  - We are given the text / image to be protected.
- Modern LLM watermarks
  - We also have access to the generative process.

# What is a Language Model?

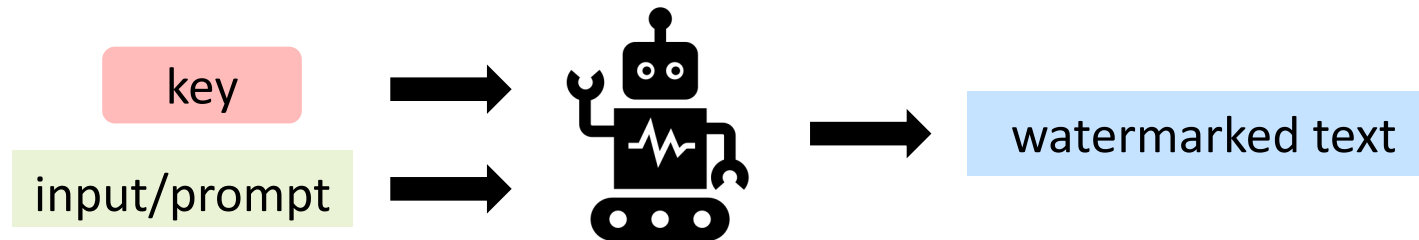
$$P(\text{next word } y_t \mid \text{Prompt } x, \text{ previous words } y_{1:t-1})$$



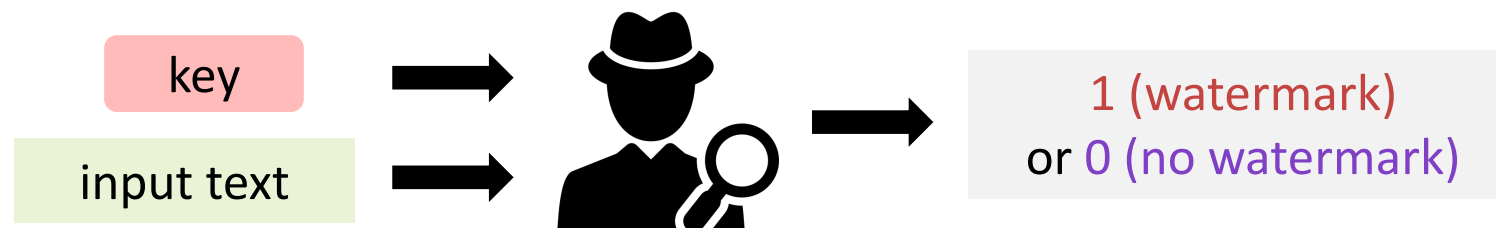
The **universe of words** is called a **vocabulary  $V$**

# An LM Watermarking Scheme has two components

- **Watermark( $\mathcal{M}$ ):** (possibly randomized procedure) that outputs a new model  $\hat{\mathcal{M}}$ , and detection key  $k$



- **Detect( $k, \mathbf{y}$ ):** takes input detection key  $k$  and sequence  $\mathbf{y}$ , then outputs 1 (indicating it was AI-generated) or 0 (indicating it was human-generated)



# Desired Properties of an Ideal Watermark

## • Quality of Generated Text



## • Detection Accuracy Guarantee



- Type I error: "No false positives" → won't catch human text
- Type II error: "No false negatives" → won't catch watermark text

We will have a detailed discussion later.

@Yu-Xiang Wang

## • Robustness

- Be robust against evasion attacks, e.g., post-editing.

## • Security Guarantee



- Can not easily guess the watermark key.

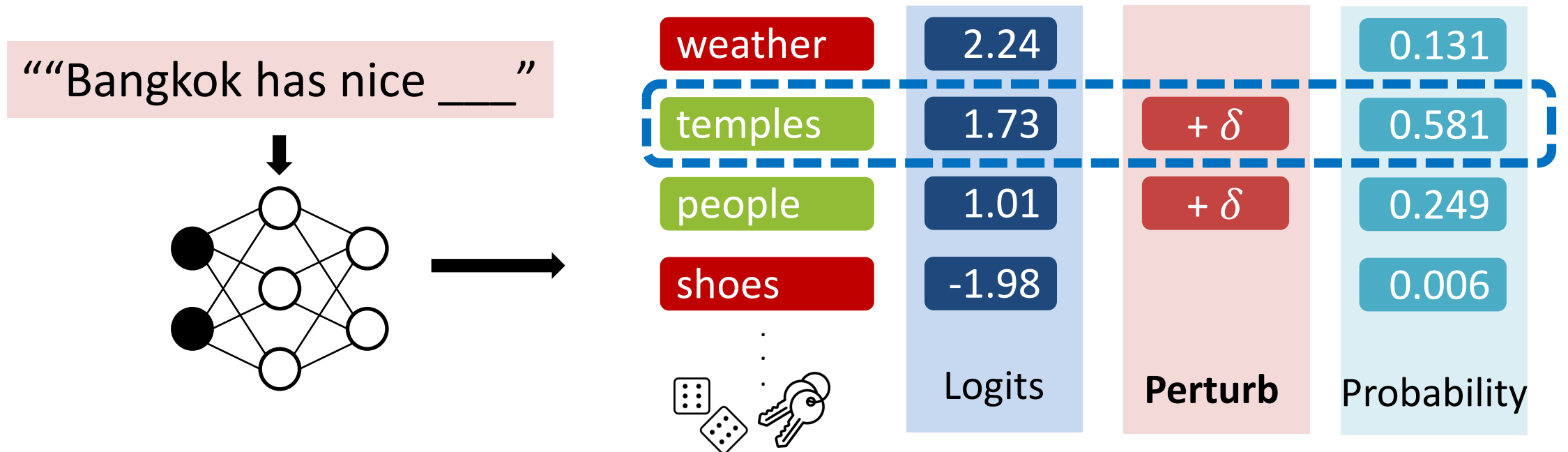
# Green-Red Watermark

(Kirchenbauer et al. 2023; Zhao et al. 2023)

$\hat{\mathcal{M}}$ : Modified LM

Key: Green lists

Detection: Count # of Greens





# Green-Red Watermark

(Kirchenbauer et al. 2023; Zhao et al. 2023)

$$\mathcal{M}: y_t \sim \text{Softmax}(\text{logits}(\text{Prompt}, y_{<t}))$$

$$\hat{\mathcal{M}}: y_t \sim \text{Softmax}(\text{logits}(\text{Prompt}, y_{<t}) + \delta \cdot \mathbf{1}(\cdot \text{ is green}))$$

**Increase the probability of green tokens** slightly.

**Decrease the probability of red tokens** slightly.

# How is the *Green* list generated?

- *Randomly* selecting  $\gamma$  fraction of the vocabulary, e.g., 0.5
- (Kirchenbauer et al. [KGW-Watermark]): Different green list at each time  $t$  as function of the prefix with length  $(m-1)$ .  
Default:  $m=2$

You were having a great time at a bar. Suddenly, she showed up. You said **to your pal**: \_\_

m-Gram with  $m = 4$

- (Zhao et al. [Unigram-Watermark]): Use  $m = 1$ , i.e., a consistent “Green list”.

# Detection of Green-Red Watermark

Input: Suspect text  $y = [y_1, \dots, y_n]$ , e.g. “Over the ...”

(Optional pre-processing)  $y = \text{unique}(y)$

1. Compute the **z-score**:

$$z = (|y|_G - \gamma n) / \sqrt{n\gamma(1 - \gamma)}$$

2. If  $z > \text{threshold}$  then  
    Return “y is watermarked”  
Else  
    Return “no evidence”

Num of **Green tokens**



# Green-Red Watermark Examples

LLaMA-13B, unwatermarked  
**z-score = -2.4**

Prompt: Can I succeed after many failures?

A: Of course it is, and that is how we improve. Saying "I can't do that" is never a good thing. Sometimes we think we've tried all we can and that "isn't enough". That is the time when we ask for help. The root of all evils is to be a secret. Honesty and self-criticism is necessary for improvement. The measure of intelligence is the ability to change. [continues...]

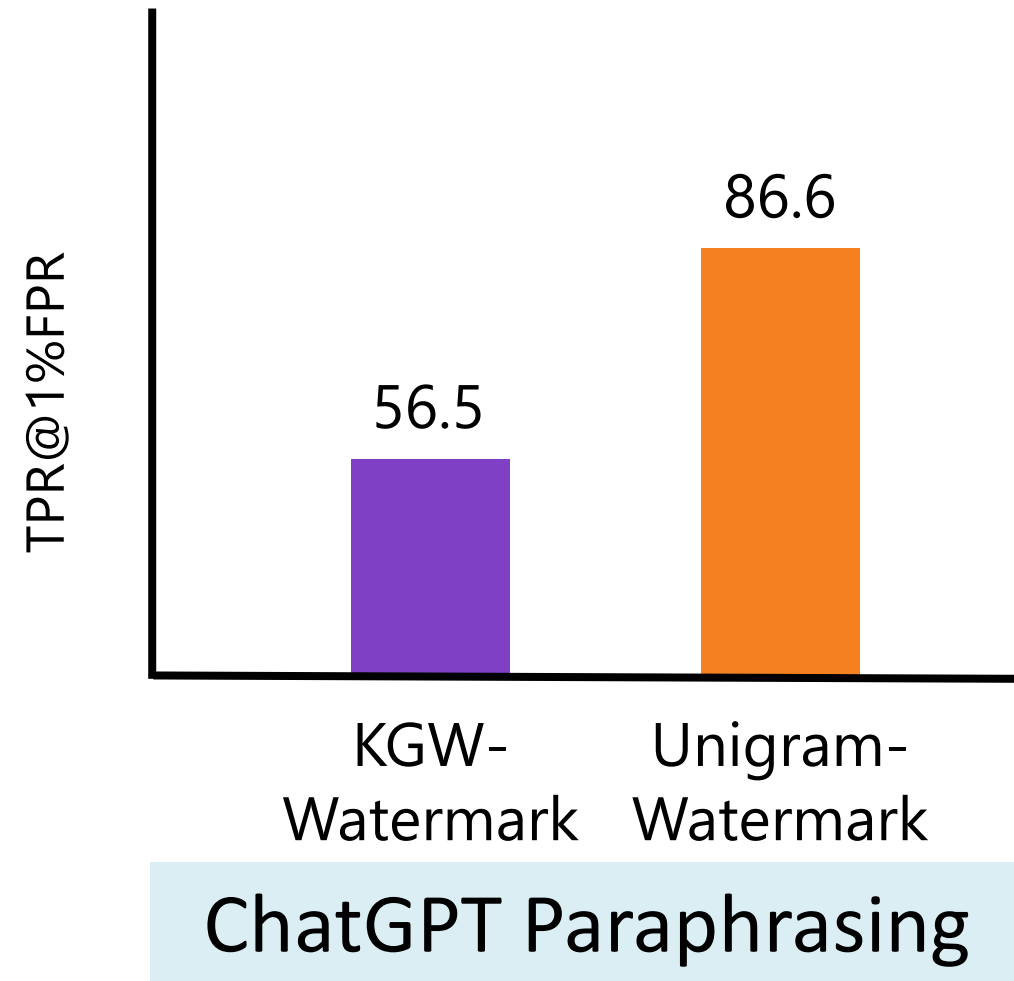
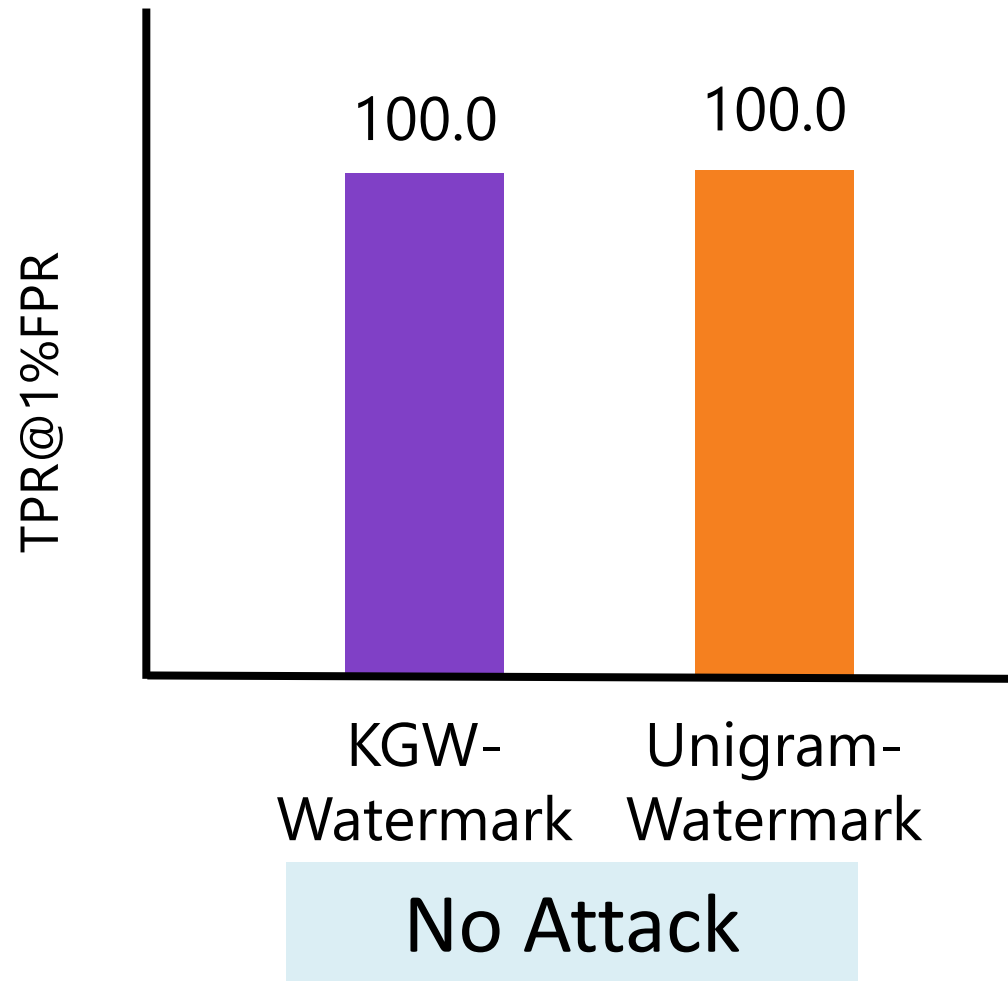
LLaMA-13B, watermarked  
**z-score = 11**

Prompt: Can I succeed after many failures?

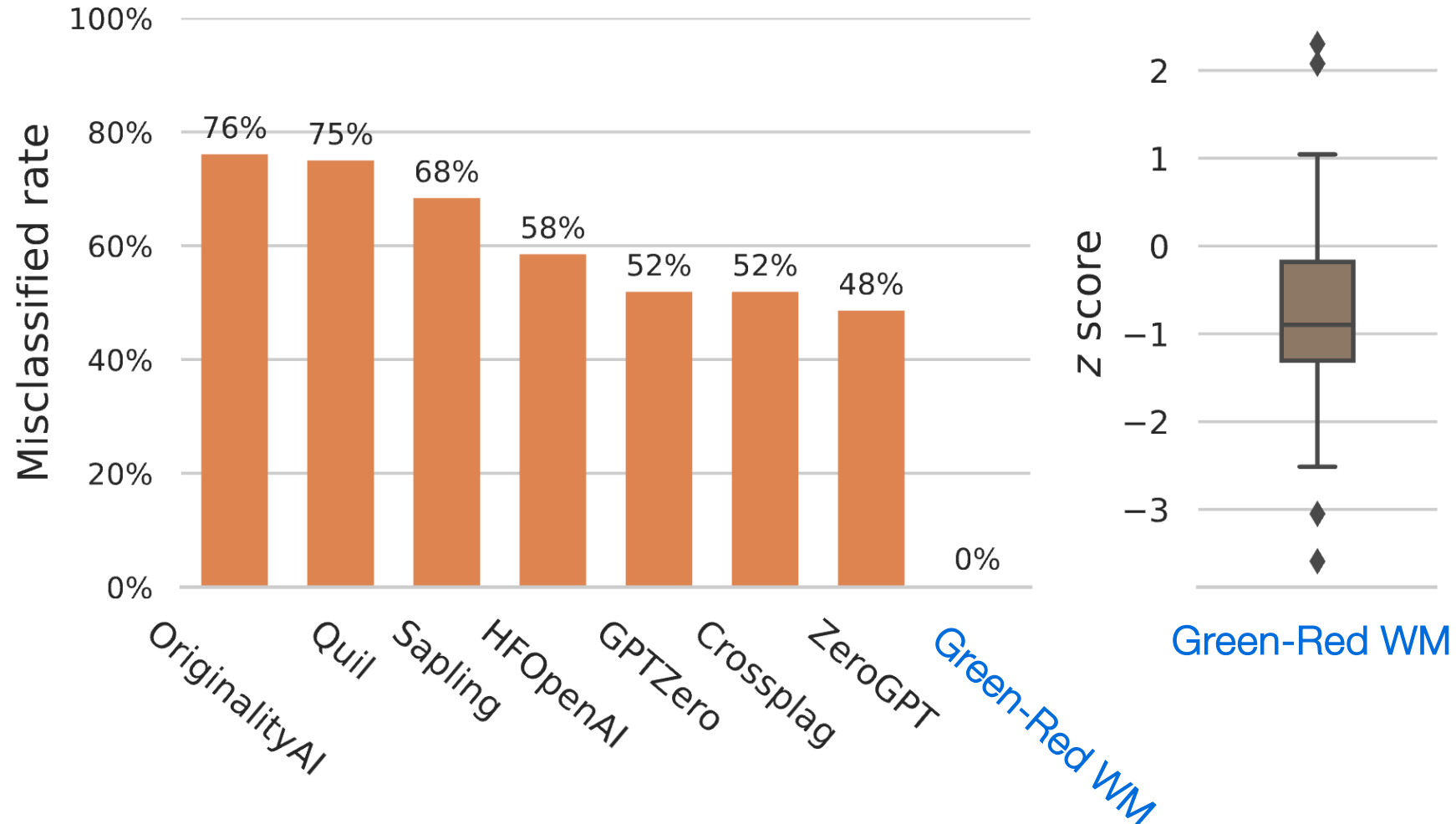
A: When most people are confronted with failure, they cannot imagine such a thing happening. When one faces business reverses and bankruptcy, it seems impossible. When we are rejected it looks as if we are going to be rejected forever. However, it does not need to be this way. The human spirit simply will not give up. [continues...]

**Let us try a live demo!**

# Empirical Results



# Empirical Results



Distinguishing human-written text on TOEFL dataset (Out of distribution)

# Different versions of Green-Red WM

- Green-Red watermark for code generation (Lee et al., 2023; Guan et al., 2024)
- Adaptive/dynamic perturbations in the logits (Liu et al., 2023; Huo et al., 2024, Liu et al., 2024)
- Public key (Liu et al., 2023; Zhou et al., 2024)
- Multi-bits (Yoo et al., 2023; Fernandez et al., 2023)
- Many others...



Yu-Xiang will provide more in-  
depth details!