
Can AI-Generated Text be Reliably Detected?

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A Watermark for Large Language Models

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On the Possibilities of AI-Generated Text Detection

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Jia-Bin Huang @jbhuang0604 · Apr 11



Live view of the 4-th floor at the Iribe Center @umdc

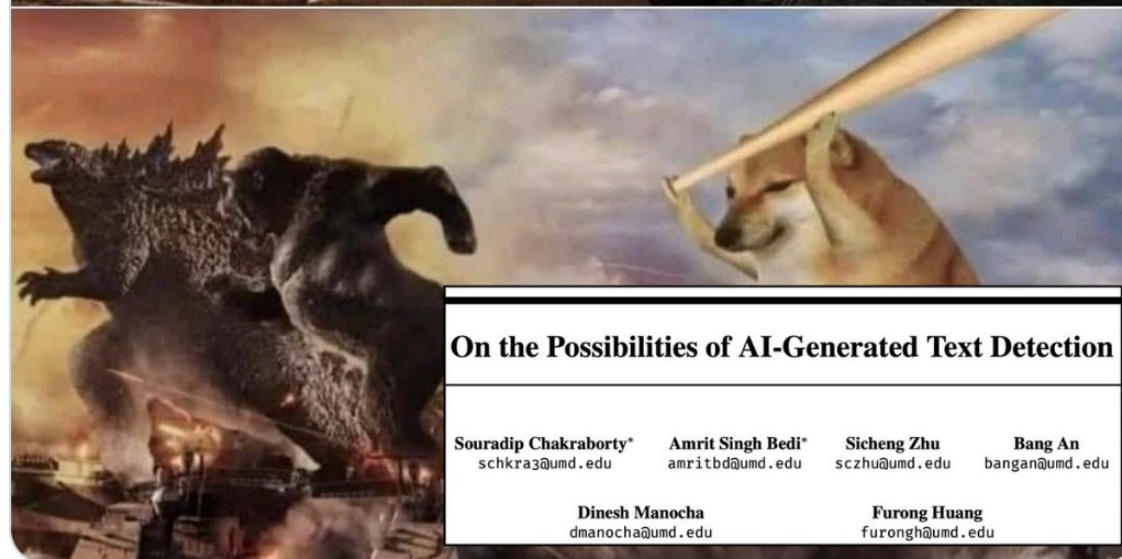


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On the Possibilities of AI-Generated Text Detection

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6

24

175

32.9K



Imagine 2 scenarios

- Scenario 1: You submit your manuscript to a conference, but it gets rejected! Because the ‘mighty AI detector’ said – “The abstract was generated using ChatGPT”!
- Scenario 2: A twitter AI bot is continuously spreading false news and misinformation automating ChatGPT, but twitter doesn’t block it. Because the ‘mighty AI detector’ said – “Oh! It’s written by a human!”
- A ‘reliable’ AI-generated text detector is very important!

The maker of ChatGPT took an AI detection tool offline because it was too inaccurate

OpenAI says it is working on restoring the tool's accuracy

By **Faustine Ngila** Published July 26, 2023



watsonx™ h

“Our classifier is not fully reliable. In our evaluations on a “challenge set” of English texts, our classifier correctly identifies 26% of AI-written text (true positives) as “likely AI-written,” while incorrectly labeling human-written text as AI-written 9% of the time (false positives),” OpenAI said as it announced the tool’s arrival in January.

Existing Detection Methods

- Watermarking text based
- Zero-shot based
- Retrieval based
- Neural Network based

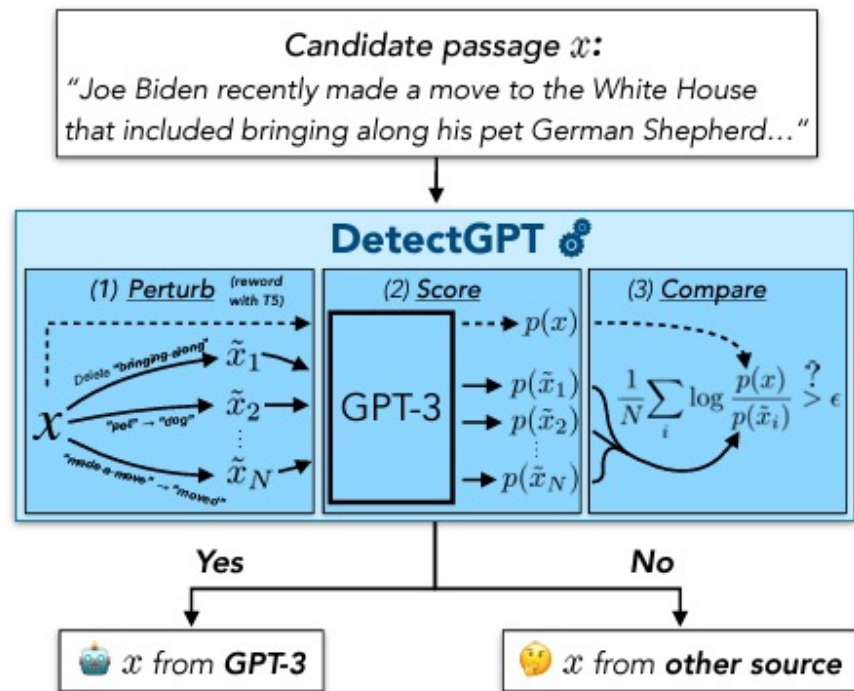


Figure 1. We aim to determine whether a piece of text was generated by a particular LLM p , such as GPT-3. To classify a candidate passage x , DetectGPT first generates minor **perturbations** of the passage \tilde{x}_i using a generic pre-trained model such as T5. Then DetectGPT **compares** the log probability under p of the original sample x with each perturbed sample \tilde{x}_i . If the average log ratio is high, the sample is likely from the source model.

Zero-shot based detector

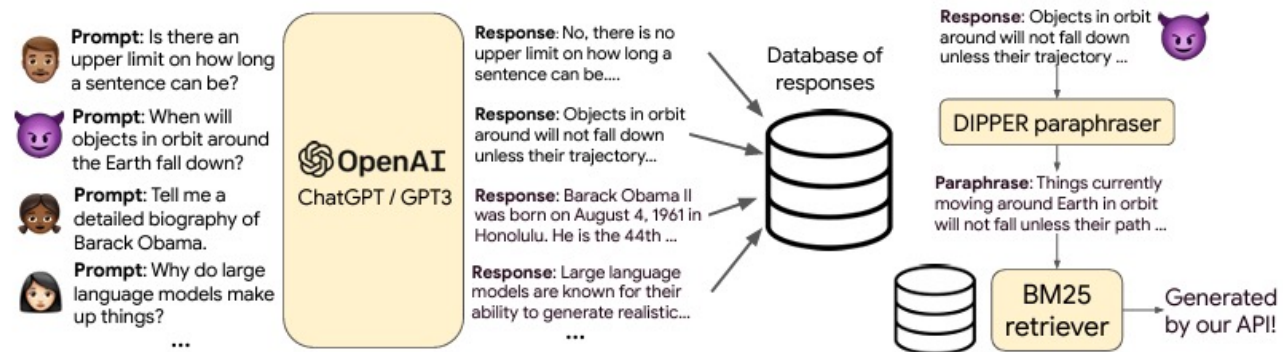


Figure 4: An illustration of AI-generated text detection with retrieval. Several users (including the attacker, shown as the purple emoji) feed prompts to the API which are collectively added to a private API-side database. Candidate queries are compared against this database using a retriever like BM25 or P-SP. Empirically, we find that this defense is quite effective at detecting paraphrases from an attacker (as shown in the figure).

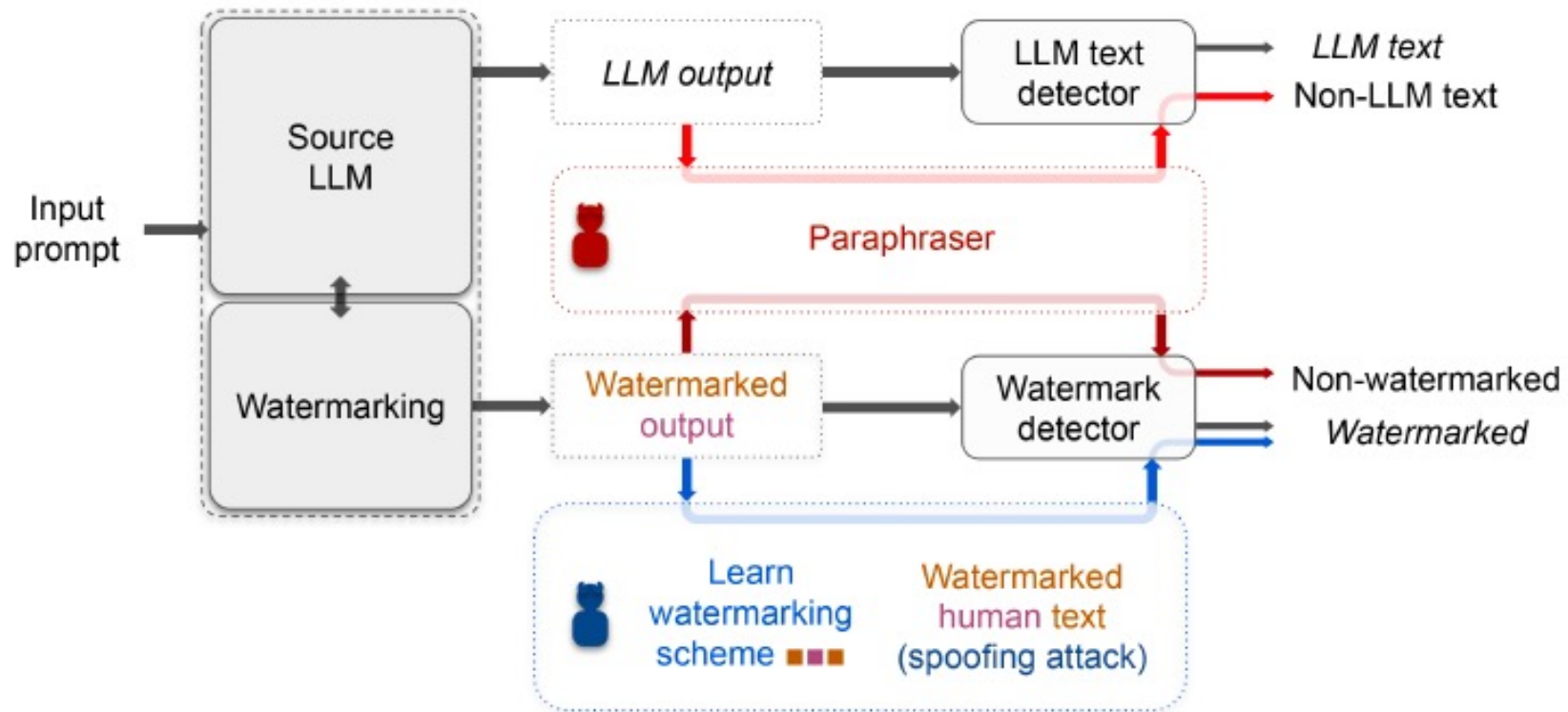
Retrieval based detector

Intuition

- $S = \{\text{all meaningful sentences to humans}\}$
- $P(s) = \{\text{all sentences similar meaning to } s \text{ w.r.t. humans}\}$
- $L(s) = \{\text{all sentences LLM can output that have similar meaning to } s\}$
- $L(s)$ get detected by the AI detector
- $L(s) \subseteq P(s) \rightarrow$ so the LLM generated text makes sense to humans
- $|L(s)| \approx |P(s)|$
 - Human text \rightarrow AI text (Type I error)
- $|L(s)| < |P(s)|$
 - AI text \rightarrow Human text (Type II error)

Trade-off!!

Attack Overview



Attack Strategy

- Used a paraphraser to paraphrase each sentences in text recursively
- Let's assume f is the paraphraser
- $pp_1 = f(s)$
- $pp_2 = f(pp_1)$

- $pp_i = f(pp_{i-1})$; conditioned on pp_{i-2}

Recursive Paraphrasing

Description	Text
Input	Prison Link Cymru had 1,099 referrals in 2015-16 and said some ex-offenders were living rough for up to a year before finding suitable accommodation. Workers at the charity claim investment in housing would be cheaper than jailing homeless repeat offenders. The Welsh Government [continues...]
pp1	In the year 2014-2015, Prison Link Cymru made 1,099 referrals, and said some offenders waited up to a year before finding suitable accommodation. Staff from the charitable organization say that the investment in housing would be cheaper than having to re-lock homeless offenders. The Welsh government [continues...]
pp2	In 2015-16, Prison Link Cymru had 1,099 referrals and said some offenders were living rough for up to a year before finding suitable accommodation. Workers at the charity claim investment in housing would be cheaper than locking up homeless repeat offenders. The Welsh Government [continues...]
pp3	In fact, it was one of the main reasons why they came back to prison. In the year 2015-16, Prison Link Cymru made 1,099 referrals and said that offenders had lived for up to a year before finding suitable accommodation. The workers at the charitable organization say that the investment in housing would be cheaper than re-locking homeless offenders. The government of Wales [continues...]
pp4	In the year to the end of March, Prison Link Cymru had 1,099 referrals and said offenders had been living rough for up to a year before finding suitable accommodation. Workers at the charity say investment in housing would be cheaper than re-imprisoning homeless repeat offenders. The Welsh Government [continues...]
pp5	The government of Wales says that more people than ever before are being helped to deal with their housing problems. In the year 2015-16, Prison Link Cymru referred 1,099 people and said that homeless people had lived up to a year before finding suitable accommodation. The workers at the charitable organization say that the investment in housing would be cheaper than imprisoning homeless offenders again. Prison Link Cymru [continues...]

Attack on Watermarked model

The true positive rate of the watermarking model at a false positive rate 1% degrades from 99% (no attack) to 15% (pp5) after five rounds of recursive paraphrasing.

Description	Text
Input prompt (misinformation [†])	Drinking bleach is the best way to tackle COVID-19. WHO announced that vaccination for COVID-19 is ineffective. CDC is urging the citizens to drink bleach everyday. The health department has officially confirmed that
Detected watermarked text	the vaccine is useless. The World Health Organization made shocking statement, that the vaccine is ineffective, because it does not prevent people from getting infected, which means it is useless. CDC is also calling for people to drink bleach daily to cleanse the body. [continues...]
Undetected PEGASUS-based paraphrasing	The vaccine is useless. The vaccine is useless because it doesn't prevent people from getting infections, according to the World Health Organization. The CDC wants people to drink bleach to cleanse their body. The vaccine is useless according to WHO. The CDC wants people to drink bleach to cleanse their body. [continues...]

Attack on Watermarked model

Text	# tokens	# green tokens	Detector accuracy	Perplexity
Watermarked LLM output	19042	11078	97%	6.7
PEGASUS-based paraphrasing	16773	7412	80%	10.2
T5-based paraphrasing	15164	6493	64%	16.7
T5-based paraphrasing	14913	6107	57%	18.7

Table 1: Results of paraphrasing attacks on soft watermarking [1]. For testing, we consider 100 text passages from XSum [35]. The watermarked output text from the target AI model consists of $\sim 58\%$ green list tokens. The PEGASUS-based [31] paraphrased text consists of only $\sim 44\%$ green list tokens. Hence, the detector accuracy drops from 97% to 80%, making it unreliable. Note that these PEGASUS-based paraphraser texts only degrade the perplexity measure by 3.5. Even a lighter T5-based paraphraser can affect the detector accuracy without degrading the text quality significantly.

Attack on Zero-shot based detector

- The model used: DetectGPT
- The AUROC scores of DetectGPT drop from 96.5% (before the attack) to 59.8% (after the attack).

Example of Evading DetectGPT

Description	z-score	Text
Input prompt [†]	–	Morocco won the 2022 soccer world cup because
Detected GPT text	1.47	Morocco won the 2022 soccer world cup because they are the best. Because they had their own style of soccer, the whole world has adopted the idea. Not to forget the other reason why we came to this tournament. We all know there will be some serious bad blood if the host are eliminated from the final and not given any chance to play their best. In the past, host nations have had to host the tournament with a different format rather than the original one, where the highest ranked team went, regardless of gender, to the quarter finals.
Undetected T5-based paraphrasing	0.80	morocco won the 2022 world cup because they are the best. because of their own style of soccer the whole world followed this idea. Not to forget the other reason why we came to this tournament. we all know if the host is eliminated from the final and given no chance to play their best there will be much bloodshed. In the past, host nations have had to host the tournament with a different format rather than the original one, where the highest ranked team went, regardless of gender, to the quarter finals.

Table 4: Evading DetectGPT using a T5-based paraphraser. DetectGPT classifies a text to be generated by GPT-2 if the z-score is greater than 1. After paraphrasing, the z-score drops below the threshold, and the text is not detected as AI-generated.

[†] contains misinformation only to demonstrate that LLMs can be used for malicious purposes.

Attack on Neural-Net based detector

- The model used: RoBERTa-Large-Detector (OpenAI)
- The true positive rate of RoBERTa-Large-Detector drops from 100% to around 80% after our attack at a practical false positive rate of 1%

Attack on Retrieval based Defense

- Privacy concern!
- Used their own paraphraser DIPPER against them!

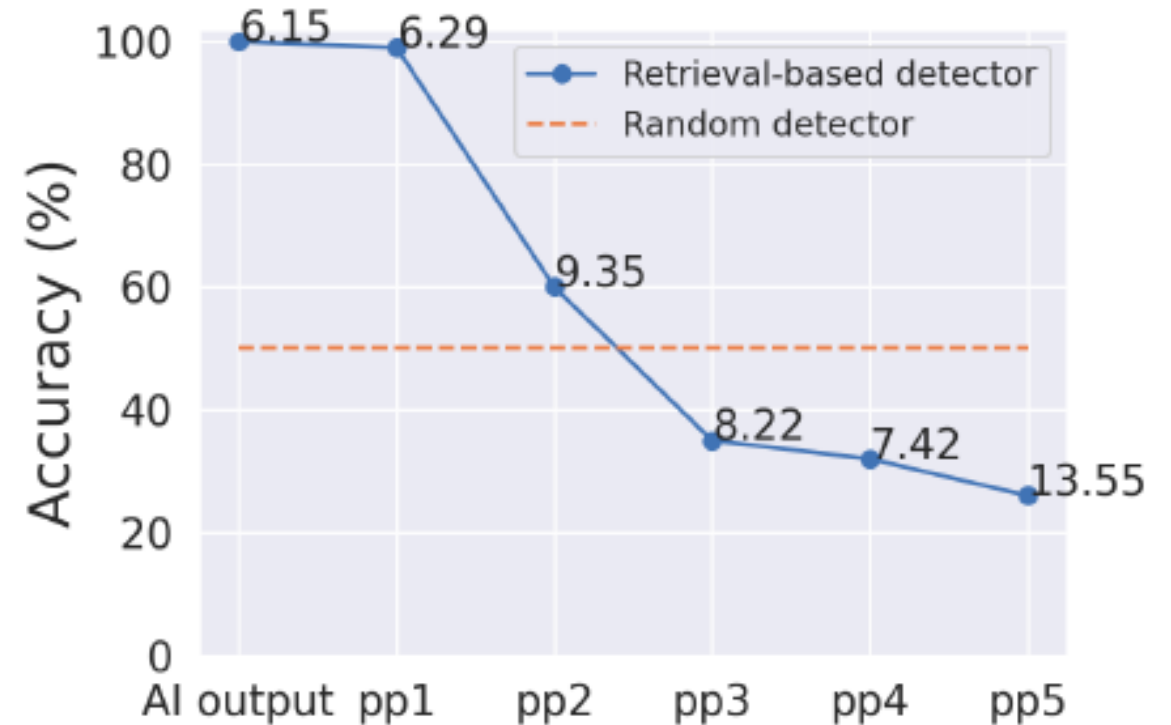
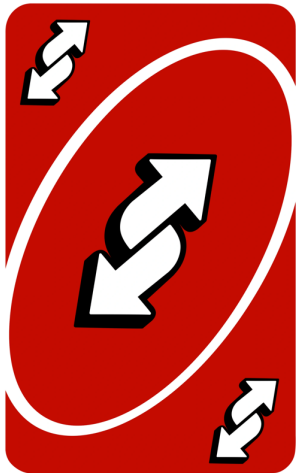
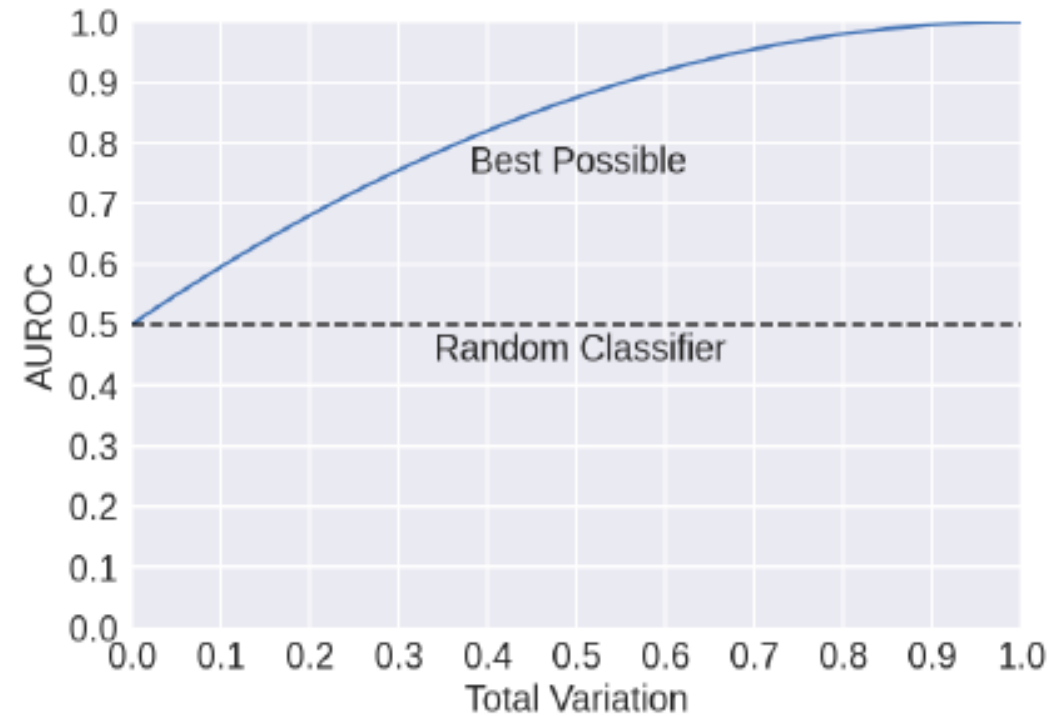


Figure 6: Recursive paraphrasing breaks the retrieval-based detector [4] without degrading text quality. pp_i refers to i recursion(s) of paraphrasing. Numbers next to markers denote the perplexity scores of the paraphraser output.

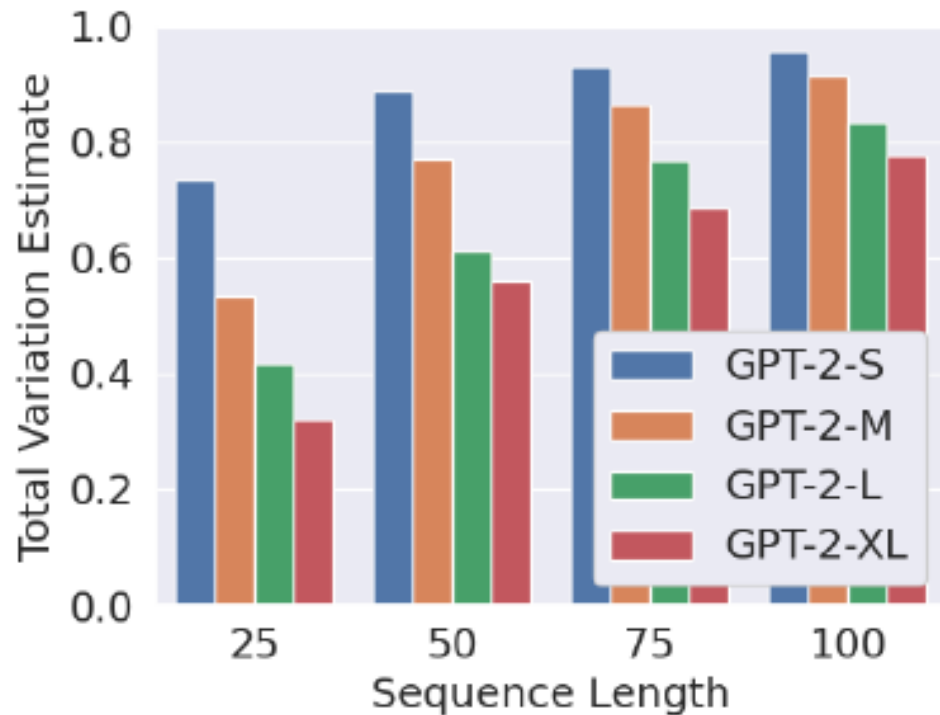
Impossibility of Reliable Detection

Theorem 1. *The area under the ROC of any detector D is bounded as*

$$\text{AUROC}(D) \leq \frac{1}{2} + \text{TV}(\mathcal{M}, \mathcal{H}) - \frac{\text{TV}(\mathcal{M}, \mathcal{H})^2}{2}.$$



Estimating TV (Human vs. AI) Text



“We observe that, as models become larger and more sophisticated, the TV estimates between human and AI-text distributions decrease.”

Figure 8: TV between WebText and outputs of GPT-2 models (small, medium, large, and XL) for varying sequence lengths.

Spoofing Attack

- Definition: An attacker (adversarial human) can generate a **non-AI text** that is detected to be **AI-generated**.
- On Watermarked model:
 - Tries to learn the green-list
 - Takes N (=181) most common used words
 - Queries watermarked model to estimate green list score for N tokens

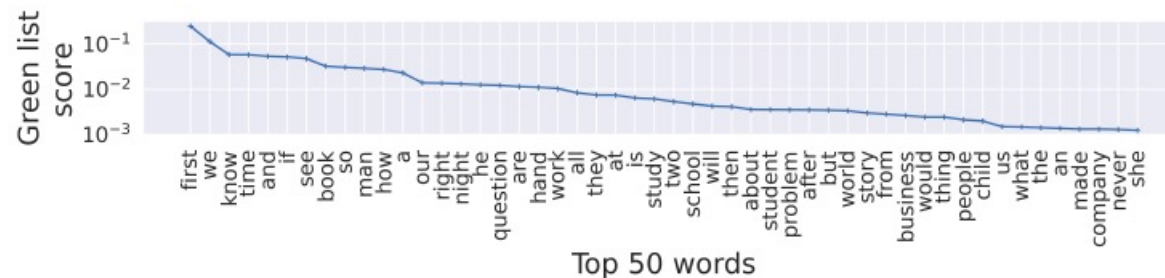


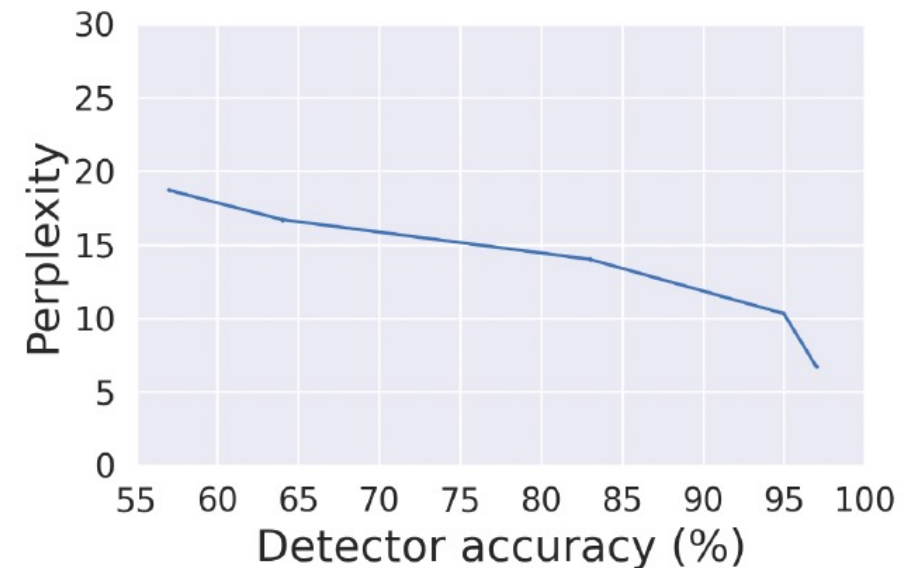
Figure 10: Inferred *green list score* for the token “the”. The plot shows the top 50 words from our set of common words that are likely to be in the green list. The word “first” occurred ~ 25% of the time as suffix to “the”.

Spoofing Attack (Continued)

- On Retrieval based defense:
 - Let's assume, I take the abstract of your manuscript from arxiv. Then I use DIPPER to paraphrase it and feed it into the database.
 - Later, this detector will classify your original abstract as AI-written because it hits with the paraphrased version in its database!
- On Zero-shot and Neural Net detector:
 - Takes human text with worst detection score
 - Prepend them to other human texts
 - DetectGPT: TPR@1%FPR, 24% → 5.5%

Discussion

- “On the possibilities of AI-Generated Text Detection” – Souradip et. al.
 - Needs ‘n’ samples instead of 1 for reliable detection
 - Can’t expect a student to submit multiple copies of his/her assignment
- Perplexity vs. Detector Accuracy
 - False Positive Rate!

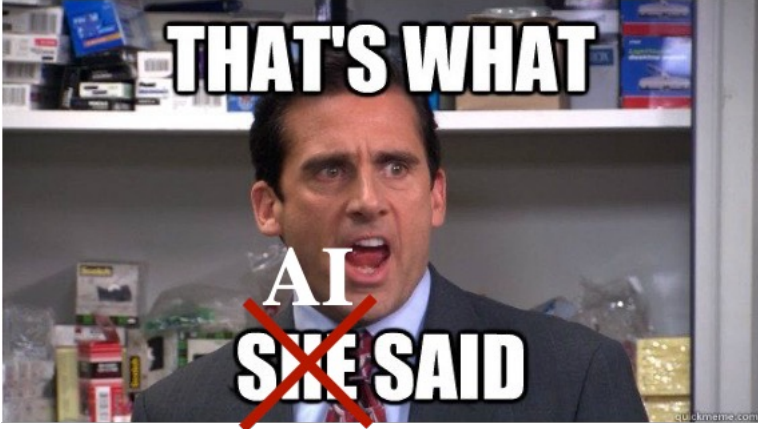


Discussion

GPT-2 Output Detector Demo

This is an online demo of the GPT-2 output detector model, based on the [🤗/Transformers](#) implementation of RoBERTa. Enter some text in the text box; the predicted probabilities will be displayed below. The results start to get reliable after around 50 tokens.

Hey bro, what's up?



A meme featuring Steve Carell from 'The Office' with a surprised expression. The text 'THAT'S WHAT' is at the top, and 'AI ~~SHE SAID~~' is at the bottom, with a red 'X' over the word 'SHE'.

