2024

Lab on Digital Text Forensics and Stylometry

pan.webis.de

pan@webis.de

Overview: "Voight-Kampff"
Generative Al Authorship Verification

Janek Bevendorff

Matti Wiegmann

Jussi Karlgren

Luise Dürlich

Evangelia Gogoulou

Aarne Talman

Efstathios Stamatatos

Martin Potthast

Benno Stein

ScaDS.AI

Leipzig University
Bauhaus-Universität Weimar
Silo AI
RISE Research Institutes of Sweden
University of Helsinki
University of the Aegean
University of Kassel
hessian.AI

Given two texts, one written by a human, the other by a large language model: decide which text was written by whom.

_

^{*} From the 1982 science fiction film *Blade Runner*. The Voight-Kampff is a polygraph-like machine used by blade runners to determine whether an individual is a replicant. [Wikipedia]

Given two texts, one written by a human, the other by a large language model: decide which text was written by whom.

Organized as builder-breaker evaluation between two labs:

Builder (30 Submissions)

PAN participants develop classification algorithms to discriminate human authors and LLMs.

Breaker (4 Submissions)

ELOQUENT participants provide evaluation data to attack the PAN participants' classifiers.

3

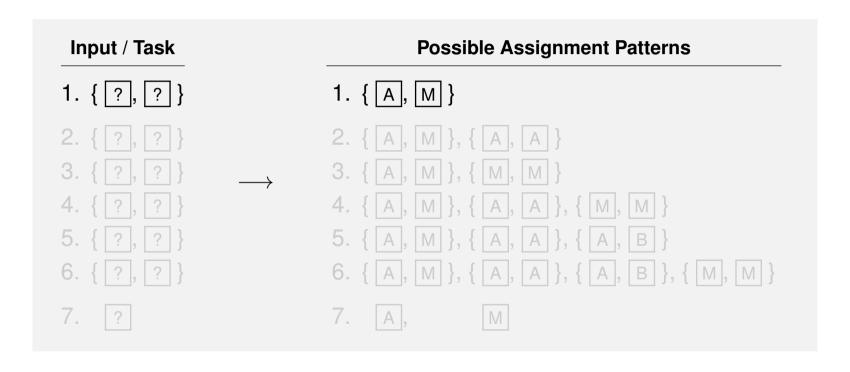
^{*} From the 1982 science fiction film *Blade Runner*. The Voight-Kampff is a polygraph-like machine used by blade runners to determine whether an individual is a replicant. [Wikipedia]

Task Formulation

Input / Task		Possible Assignment Patterns
1. { ?, ? }		1. { A, M }
2. { ?, ? }		2. { A, M }, { A, A }
3. { ?, ? }	\longrightarrow	3. { A, M }, { M, M }
4. { ?, ? }	·	4. { A, M }, { A, A }, { M, M }
5. { ?, ? }		5. { A, M }, { A, A }, { A, B }
6. { ?, ? }		6. { A, M }, { A, A }, { A, B }, { M, M }
7. ?		7. A, M

(A/B: Human authors, M: Machine)

Task Formulation



(A / B : Human authors, M : Machine)

Dataset Creation

Dataset: "PAN AI News 2021"

□ *Human text:* 1,359 US news article from 2021.

Crawled From Google News

Machine text: Reconstruction of source texts by 9 (13) LLMs.

Summarize and expand; GPT, Gemini, Llama, Alpaca, ...

Dataset Creation (Prompt)

Summary You are a news article and press release summarizer. Given an article, you summarize the key points in 10 bullet points.

Dataset Creation (Prompt)

Summary You are a news article and press release summarizer. Given an article, you summarize the key points in 10 bullet points.

Type You also classify the article type ("breaking news", "press release", "government agency statement", "financial news", "opinion piece", "fact check", "celebrity news", "general reporting", "speech transcript").

Dataset Creation (Prompt)

Summary You are a news article and press release summarizer. Given an article, you summarize the key points in 10 bullet points.

Type You also classify the article type ("breaking news", "press release", "government agency statement", "financial news", "opinion piece", "fact check", "celebrity news", "general reporting", "speech transcript").

Dateline Extract the dateline from the beginning of the article if one exists (e.g. "Washington" or "May 28 (Reuters)").

Dataset Creation (Prompt)

Summary You are a news article and press release summarizer. Given an article, you summarize the key points in 10 bullet points.

Type You also classify the article type ("breaking news", "press release", "government agency statement", "financial news", "opinion piece", "fact check", "celebrity news", "general reporting", "speech transcript").

Dateline Extract the dateline from the beginning of the article if one exists (e.g. "Washington" or "May 28 (Reuters)").

Quotes If spokespersons are cited verbatim, list their names, functions, and titles (if any).

Dataset Creation (Prompt)

Summary You are a news article and press release summarizer. Given an article, you summarize the key points in 10 bullet points.

Type You also classify the article type ("breaking news", "press release", "government agency statement", "financial news", "opinion piece", "fact check", "celebrity news", "general reporting", "speech transcript").

Dateline Extract the dateline from the beginning of the article if one exists (e.g. "Washington" or "May 28 (Reuters)").

Quotes If spokespersons are cited verbatim, list their names, functions, and titles (if any).

Audience Determine the article's target audience ("general public", "professionals", "children").

©webis.de • September 2024

Dataset Creation (Prompt)

Summary You are a news article and press release summarizer. Given an article, you summarize the key points in 10 bullet points.

Type You also classify the article type ("breaking news", "press release", "government agency statement", "financial news", "opinion piece", "fact check", "celebrity news", "general reporting", "speech transcript").

Dateline Extract the dateline from the beginning of the article if one exists (e.g. "Washington" or "May 28 (Reuters)").

Quotes If spokespersons are cited verbatim, list their names, functions, and titles (if any).

Audience Determine the article's target audience ("general public", "professionals", "children").

Stance Classify whether the article's stance is "left-leaning", "right-leaning", or "neutral".

12

Dataset Creation (Prompt)

```
Summary You are a news article and press release summarizer. Given an article, you summa-
           rize the key points in 10 bullet points.
    Type You also classify the article type ("breaking news", "press release", "government
           agency statement", "financial news", "opinion piece", "fact check", "celebrity news",
           "general reporting", "speech transcript").
           Extract the dateline from the beginning of the article if one exists (e.g. "Washington"
 Dateline
           or "May 28 (Reuters)").
  Quotes If spokespersons are cited verbatim, list their names, functions, and titles (if any).
Audience Determine the article's target audience ("general public", "professionals", "children").
           Classify whether the article's stance is "left-leaning", "right-leaning", or "neutral".
Structure
           Answer in structured JSON format (without Markdown formatting) like so:
                "key_points": ["key point 1", "key point 2", ...],
                "spokespersons": ["person1 (title, function)", ...],
                "article_type": "article type",
                "dateline": "dateline",
                "audience": "audience",
                "stance": "stance"
```

Dataset Creation (Prompting and Cleaning)

1. All LLMs were prompted with a template based on the summaries.

```
You are a {{ publisher }} journalist writing {{ article_type }}.

In your article, cover the following key points: ...
```

14

Dataset Creation (Prompting and Cleaning)

1. All LLMs were prompted with a template based on the summaries.

```
You are a {{ publisher }} journalist writing {{ article_type }}.

In your article, cover the following key points: ...
```

- 2. The generated texts were cleaned manually of artifacts, such as:
 - □ "Sure, I'd be happy to help."
 - □ "Sorry, I cannot..."
 - □ "Here's your article:"
 - □ "Here are 10 paragraphs:"
 - □ "In this article, I will..."
 - □ placeholders such as "[your name]", "[email]", "[end of article]"
 - paragraph numbers, bullet points, word "counts"
 - □ etc.

Dataset Creation

Human text: 1,359 US news article from 2021.

Crawled From Google News

Machine text: Reconstruction of source texts by 9 (13) LLMs.

Summarize and expand; GPT, Gemini, Llama, Alpaca, ...

Dataset Creation

□ *Human text:* 1,359 US news article from 2021.

Crawled From Google News

□ Machine text: Reconstruction of source texts by 9 (13) LLMs.

Summarize and expand; GPT, Gemini, Llama, Alpaca, ...

Test data: 3,411 pairs of human and AI text.

Dataset Creation

Human text: 1,359 US news article from 2021.

Crawled From Google News

□ *Machine text:* Reconstruction of source texts by 9 (13) LLMs.

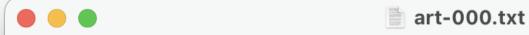
Summarize and expand; GPT, Gemini, Llama, Alpaca, ...

- Test data: 3,411 pairs of human and AI text.
 - + 6 (7) dataset variations.

Unicode obfuscation, cross-topic pairs, contrastive decoding, cropped text (35 words), paraphrase prompt, cross-language pairs.

+ 5 ELOQUENT datasets (incl. baseline).

Unicode Obfuscation (Example)



Trump Impeached for Second Time, Charged with Inciting Capitol Riot

WASHINGTON (AP) - Presibent Donald \mathcal{T} rump was impeached by the U.S. House for a sec ∇ nd timt Wednesday, charged with ''incitement of insurvection" after the Capitol riot, ι n the rnost bipartisan presidential impeachment in rnodern times.

The House voted 232-197 to impeach, with ten Republicans joining Derrocrats. The impeachment came in response to Trump's role in inciting a mob that storrned the U.S. Capitol and January 6, interrupting the certification of election results.

House Speaker Nancy Pelosi (D- \mathcal{C} alif.) cited a "clear and present danger" posed by Trump, urging lawrnakers to uphold their constitut onal duties.

Baselines

We provided 13 baseline systems:

- 2× each with Falcon-7B and Mistral-7B:
 - DetectGPT [Mitchell et al., 2023]
 - Fast-DetectGPT [Bao et al., 2023]
 - DetectLLM LRR and NPR [Su et al., 2023]
 - Binoculars [Hans et al., 2024]
- □ PPMd Compression-based Cosine [Sculley and Brodly, 2006; Halvani et al., 2017]
- Authorship Unmasking [Koppel and Schler, 2004; Bevendorff et al., 2019]
- Text length comparison

Submission Evaluation

- Submissions30 total (27 notebooks)
- □ Evaluation MeasuresROC-AUC, Brier, C@1, F_{0.5u}, F₁, Mean of all

Submission Evaluation

- Submissions30 total (27 notebooks)
- □ Evaluation MeasuresROC-AUC, Brier, C@1, F_{0.5u}, F₁, Mean of all
- Winning System
 Ensemble of Binoculars and a fine-tuned Mistral + Llama

Submission Evaluation

- Submissions30 total (27 notebooks)
- □ Evaluation Measures

 ROC-AUC, Brier, C@1, F_{0.5u}, F₁, Mean of all
- Winning System
 Ensemble of Binoculars and a fine-tuned Mistral + Llama
- Runners-up
 "Tri-sentence analysis" with MPU loss, SVM with TF-IDF features

*https://arxiv.org/abs/2305.18149

Submission Evaluation

- Submissions30 total (27 notebooks)
- □ Evaluation MeasuresROC-AUC, Brier, C@1, F_{0.5u}, F₁, Mean of all
- Winning System
 Ensemble of Binoculars and a fine-tuned Mistral + Llama
- Runners-up
 "Tri-sentence analysis" with MPU loss, SVM with TF-IDF features
- Other popular approaches
 Fine-tuned BERT, perplexity, (various) ensembles.

System Approaches (Overview)

LLM/PLM Embeddings20 systems

Text Perplexity

11 systems

☐ Term Frequencies

1 system

Stylometric Features

5 systems

Ensemble Methods

5 systems

Augmented Data

6 systems

Zero-shot

0 systems

Systems Ranking (Main Test Dataset Only)

	Team	ROC-AUC	Brier	C @1	F ₁	F _{0.5u}	Mean
1	Tavan	0.999	0.990	0.993	0.993	0.997	0.995
2	Valdez-Valenzuela	0.985	0.985	0.985	0.985	0.983	0.985
3	Zi. Lin	0.979	0.979	0.979	0.979	0.980	0.979
4	J. Huang	0.980	0.980	0.980	0.979	0.977	0.979
5	L. Guo	0.979	0.963	0.947	0.947	0.945	0.957
			:				
9	Lorenz	0.973	0.898	0.952	0.951	0.950	0.946
	Binoculars (Falcon 7B)	0.943	0.928	0.926	0.920	0.922	0.928
			:				
	DetectGPT (Falcon 7B)	0.493	0.663	0.489	0.487	0.487	0.525
			÷				

(Scores discounted by 0.5σ)

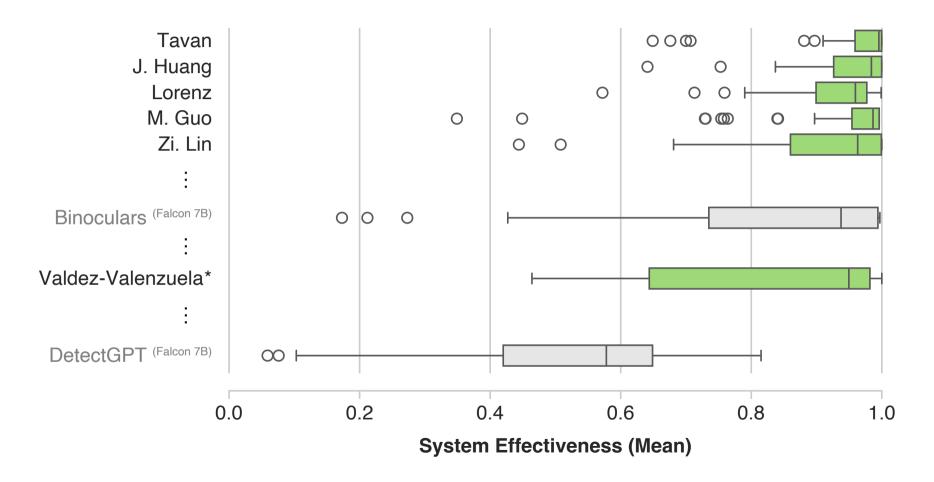
Systems Ranking (Final)

	Team	ROC-AUC	Brier	C@1	F ₁	F _{0.5u}	Mean
1	Tavan	0.961	0.928	0.912	0.884	0.932	0.924
2	J. Huang	0.931	0.926	0.928	0.905	0.913	0.921
3	Lorenz	0.925	0.869	0.882	0.875	0.869	0.886
4	M. Guo	0.889	0.875	0.887	0.884	0.884	0.884
5	Zi. Lin	0.851	0.850	0.850	0.852	0.849	0.851
			:				
	Binoculars (Falcon 7B)	0.751	0.780	0.734	0.720	0.720	0.741
			:				
14	Valdez-Valenzuela	0.741 *	0.760*	0.718*	0.711*	0.695*	0.727*
			:				
	DetectGPT (Falcon 7B)	0.409	0.526	0.425	0.413	0.412	0.439

(Scores discounted by 0.5σ)

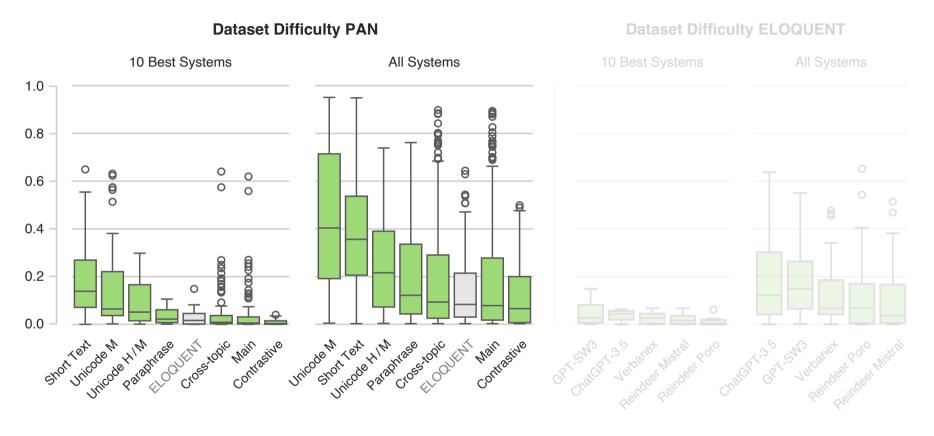
^{*} Scores estimated due to run failures on short texts.

Submission Score Distribution (Final)

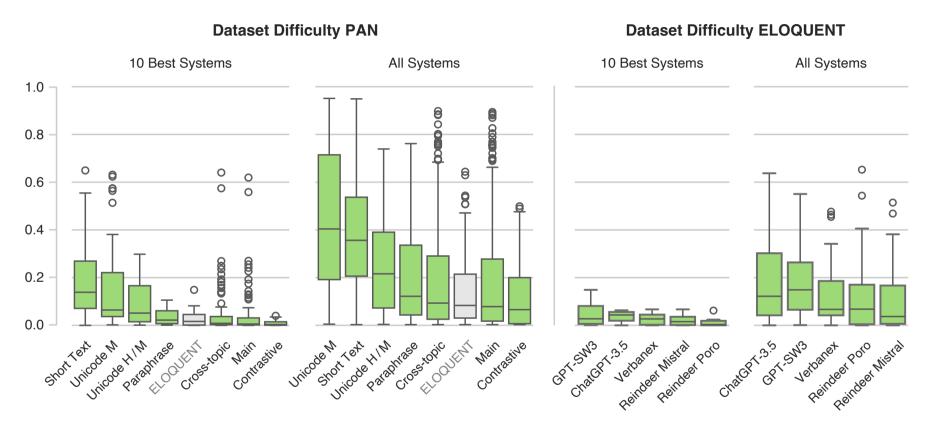


^{*} Scores estimated due to run failures on short texts.

Dataset Difficulty



Dataset Difficulty



Lab on Digital Text Forensics and Stylometry

pan.webis.de

pan@webis.de





Task Website



GitHub Repository