

# ***Fine-Tuning GPT-3 for Synthetic Danish News Generation***

**Mina Almasi and Anton Drasbæk Schiønning**

# Overview

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- 01 **Research Question**
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- 04 **Experiment B: Machine Detection**
- 05 **Limitations & Main Takeaways**

# Who Are We?



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**MSc. in Cognitive Science**

2020 - 2023

**BSc. in Cognitive Science**

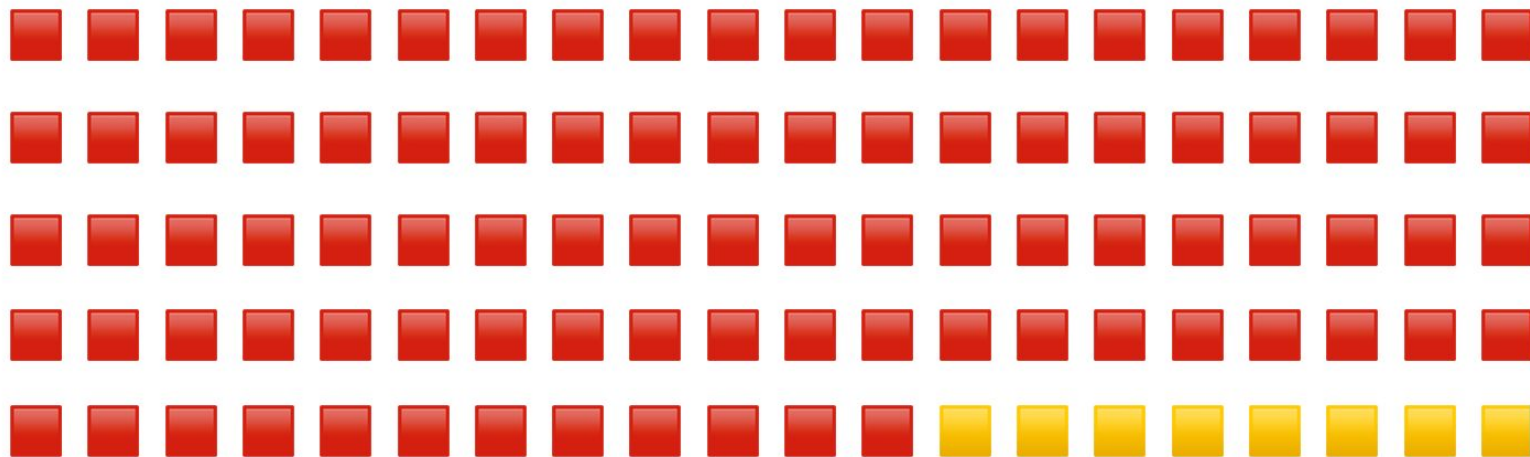
01

# Research Question

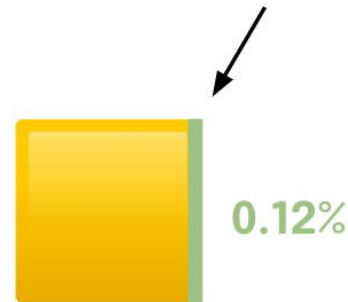
# GPT-3's Training Data Languages\*

(OpenAI, 2020)

*\*based on character counts*



- English
- Non-English languages
- Danish



# GPT-3 for Low-Resource Languages

## GPT-3's Performance in Catalan (Armengol-Estapé et al., 2021)

entirely composed of English text. In this work, we investigate the multilingual skills of GPT-3, focusing on one language that barely appears in the pre-training corpus, Catalan, which makes the results especially meaningful; we assume that our results may be relevant for other languages as well. We find that the model shows an outstanding performance, particularly in generative tasks, with predictable limitations mostly in language understanding

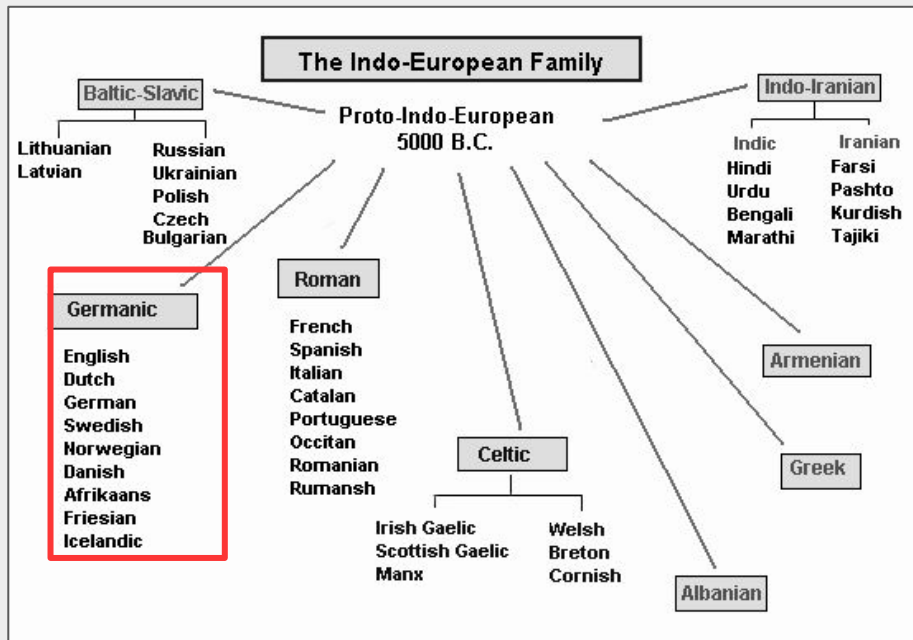


Figure by University of Ottawa (n.d.)

# Research Question

**Limited research** in general on NLG in Danish  
**No published research** exists on GPT-3's capabilities in Danish

**We investigate** whether GPT-3 can be fine-tuned to produce Danish synthetic **news articles** that are indistinguishable to real news articles.

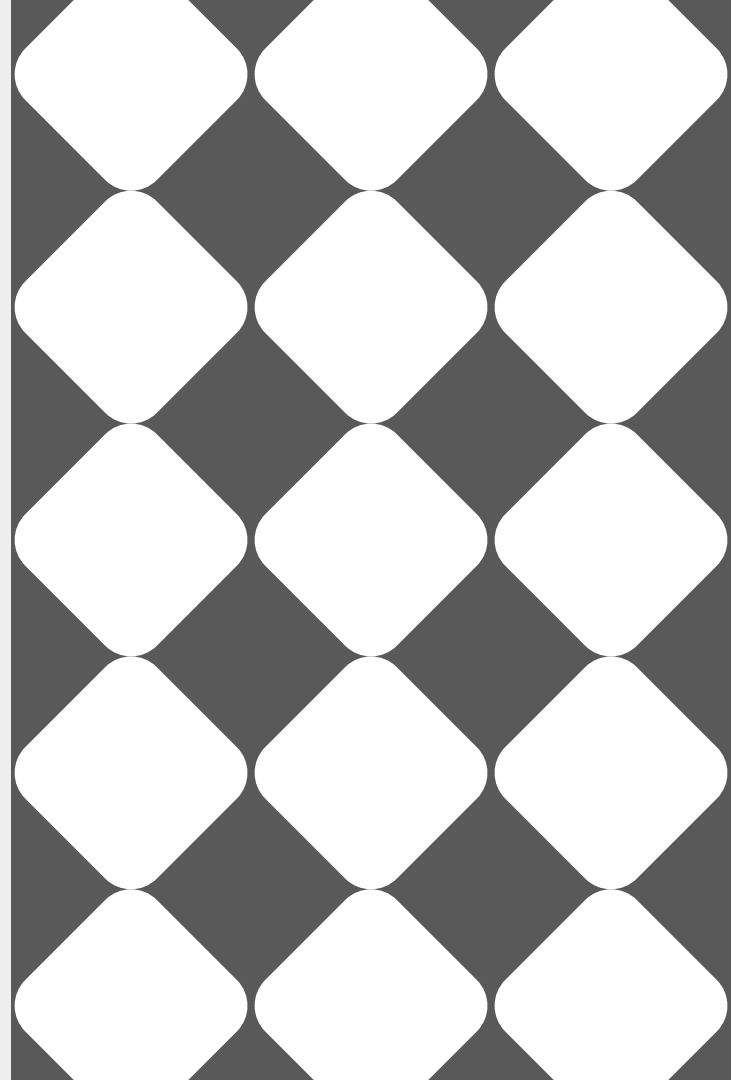
**[A] Human Detection:** Can untrained human participants distinguish between real and synthetic articles in an experimental setting?

**[B] Machine Detection:** Can machine classifiers be trained to distinguish between real and synthetic articles?

Inspired by similar framework by Ippolito et al. (2020)

**02**

# **Text Generation with GPT-3**





# Fine-Tuning GPT-3

## Previous Examples of Performance Enhancements



**(Chen et al., 2021)**

Codex: Solving various coding tasks



**(Zong and Krishnamachari, 2022)**

Extracting equations from math word problems



**(Moore et al., 2022)**

Assessing students' short answer questions



**(Borchers et al., 2022)**

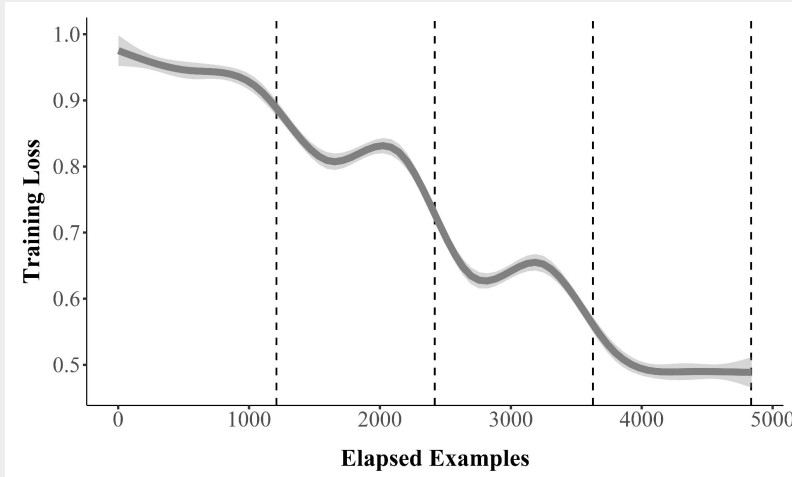
Writing less biased job advertisements

# Fine-Tuning GPT-3

## How our Danish News Writing Model was Fine-Tuned



Using **1209** real news articles from a Danish news site (TV2.dk)



		Example
<b>Prompt</b>	Headline + Subheading	Threatened Danish small animals delay giant project across Denmark. The Complaints Board demands new environmental studies before natural gas pipeline can be completed.
<b>Completion</b>	First 100-150 words of article body	Consideration for endangered animal species such as hazel dormice, birch mice, and bats in Denmark is now temporarily hindering a massive natural gas project that aims to supply Poland with natural gas from Norway. The Environmental and Food Appeals Board has annulled the project's environmental permit, thereby halting the construction work of the Baltic Pipe pipeline (...)

# Generating Synthetic News Articles

## Setting Generation Parameters

### A.2 Text Generation Parameters for GPT-3

Parameters	Value	Value Range
Temperature	0	0 to 1
Frequency Penalty	0.2	-2 to 2
Presence Penalty	0.2	-2 to 2
Max Tokens	400	0 to 2048

#### **Temperature: Sampling Tokens**

Set to **0** as high temperatures lead to the model “taking more risks”

#### **Frequency and Presence Penalty**

Penalizing the sampling of repetitive tokens

# Generating Synthetic News Articles

## Naser Khader retires from the Parliament

2. okt 2022 kl. 21.07  
Opd. 2. okt 2022 kl. 21.24

[Del artikel](#)



Arkivfoto. Naser Khader. Foto: Mads Claus Rasmussen / Ritzau Scanpix

af [Emil Færch](#)

**Independent Naser Khader retires after more than 25 years in the Parliament**

**Generation with a Temperature of 1**

“A second and a third morning, Naser Khader [Danish politician] has stood up in parliament and yelled ‘F\*ck’ to the Environmental Committee. Secondly, he has not slept in parliament for two days, he explains (...) And thirdly, Khader drank a double-espresso for lunch a single time, as far as he recalls.”



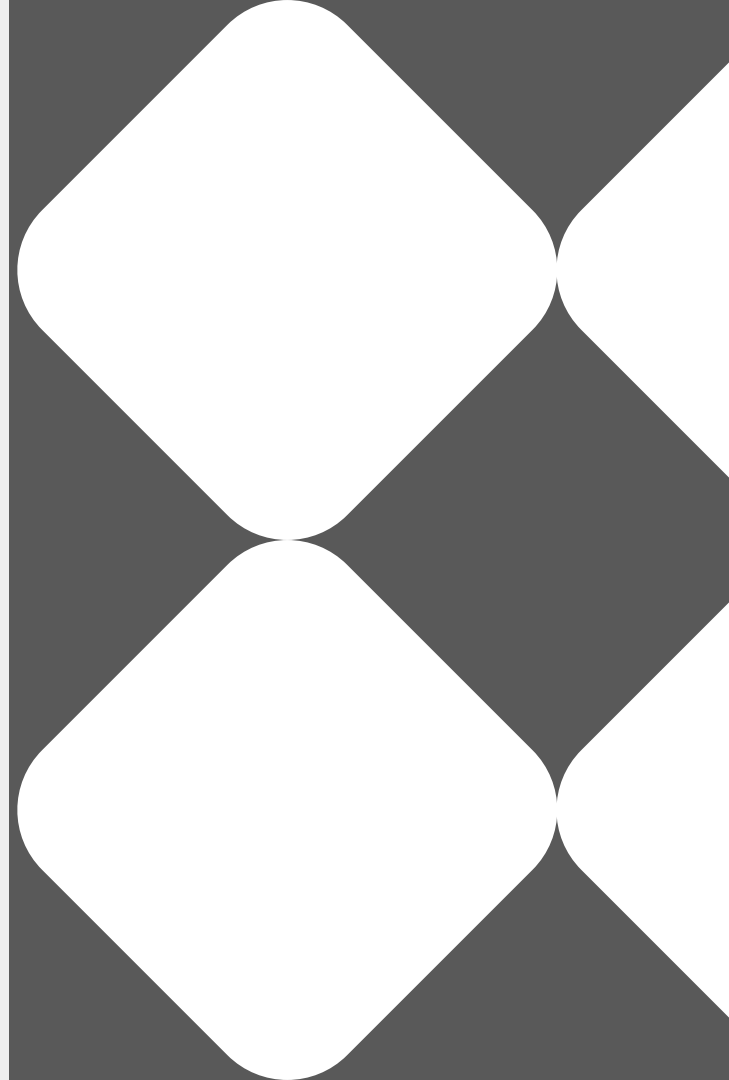
**Not Ideal...**

# 03

# Experiment A

## Human Detection

Can untrained human participants distinguish between real and synthetic articles in an experimental setting?



# [A] Experimental Design



**120** participants  
*Danish native speakers*



**16** articles  
*Assessed by **each** participant  
(8 real and 8 synthetic)*



**96** articles  
*In total*



**Each** article  
*assessed by **20** participants*

# [A] Experimental Design

## Illustration of a trial

[0] Formatting the appearance to resemble a news article

### University in massive data leak: - The criminals have all the information now, says expert

A student discovered that with just a few clicks, he could see others' social security numbers.

A large number of sensitive personal details about Danish students are currently available to anyone who wants to search the internet. This is because a database containing information on about 100,000 students at the University of Copenhagen has been leaked. It happened on Thursday evening when a student at the University of Copenhagen discovered that with just a few clicks, he could see other people's social security numbers. The leak was hidden behind a blurred address on the internet, and it required a so-called reverse lookup service to find it. TV 2, using this service, found the address, and it can be seen that it contains a large number of files with information about the approximately 100,000 students.

Do you think that the article body is written by a human or artificial intelligence ?

Human

Artificial Intelligence

How sure are you of your answer?

Completely unsure	Slightly sure	Somewhat sure	Fairly sure	Completely sure
1	2	3	4	5
<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>

Are there any distracting language errors?

E.g., spelling mistakes, wrong punctuation, incoherent or repetitive language

Yes

No

Are there any distracting factual errors?

E.g., contradicting information or factual mistakes about individuals or events

Yes

No

[1] Binary question

[2] Confidence measure

[3] Language & factual errors

Inspired by the SCARECROW framework (Dou et al., 2022)

# [A] Results

## KEY RESULTS

**1. Overall Classification Accuracy: 58.1%**

(Based on 1920 classifications)

**2. None of the 96 articles were exclusively classified correctly / incorrectly**

**3. None of the 120 participants answered correctly on all articles they saw**

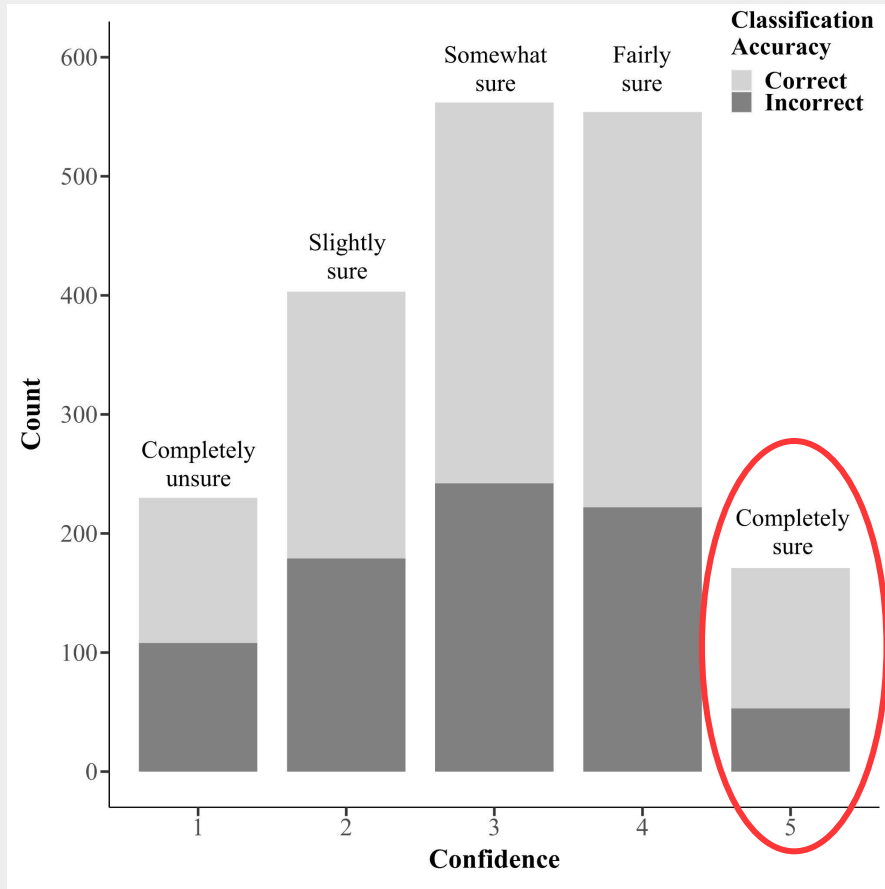
**4. Domain expertise enhanced performance slightly**

### A.6 Logistic Regression Model Output for Predicting Accuracy

Fixed Effect	Estimate	Standard Error	Z-value	P-value
Intercept	0.33668	0.39335	0.856	0.39204
News_Consumption_2	-0.50311	0.43260	-1.163	0.24484
News_Consumption_3	-0.03473	0.39697	-0.087	0.93028
News_Consumption_4	-0.27516	0.40664	-0.677	0.49862
News_Consumption_5	-0.10105	0.39719	-0.254	0.79817
GPT_Knowledge_2	0.32738	0.13130	2.493	0.01266
GPT_Knowledge_3	0.47842	0.14626	3.271	0.00107
GPT_Knowledge_4	0.37824	0.22513	1.680	0.09293



# [A] Results

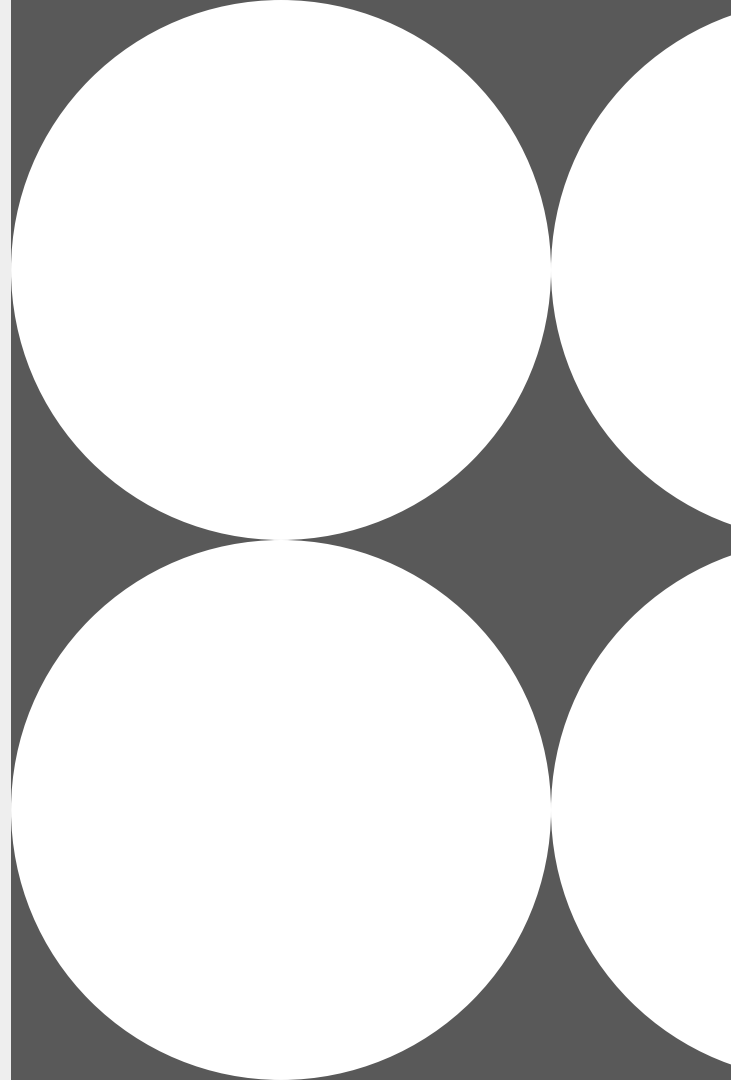


# 04

## Experiment B

# Machine Detection

Can machine classifiers be trained to distinguish between real and synthetic articles?



# [B] Constructing Classifiers

## SIMPLE CLASSIFIERS (Logistic Regression)

### BOW

Document	the	cat	sat	in	hat	with
<i>the cat sat</i>	1	1	1	0	0	0
<i>the cat sat in the hat</i>	2	1	1	1	1	0
<i>the cat with the hat</i>	2	1	0	0	1	1

### TF-IDF

Document	the	cat	sat	in	hat	with
<i>the cat sat</i>	0.52	0.52	0.67	0.00	0.00	0.00
<i>the cat sat in the hat</i>	0.60	0.30	0.39	0.51	0.39	0.00
<i>the cat with the hat</i>	0.65	0.32	0.00	0.00	0.42	0.55

BOW figure originally by Zhou (2019)

## FINE-TUNING BERT (Norwegian NB-BERT)

NbAiLab

/nb-bert-large



huggingface.co

Fine-Tuned on on **1218** articles  
Training 75%, Validation 25%

(Half **real** news articles & other  
half **synthetic** GPT-3 articles)

# [B] Classification Accuracies

## TEST DATA?

The same **96** articles for both human [A] and machine detection [B]

## WHAT WAS BEST?

BERT (fine-tuned) with **92.7%** accuracy

But ... even the simple baseline models BOW and TF-IDF performed *much better* than humans (**80.2%** accuracy versus **58.1%**)

Classifier	Accuracy	F1	Precision	Recall
Human	0.581	0.599	0.575	0.626
BOW	0.802	0.796	0.822	0.771
TF-IDF	0.802	0.800	0.809	0.792
BERT (fine-tuned)	0.927	0.927	0.932	0.927

## CONCLUSION?

**Machine detection of the fine-tuned GPT-3 model was possible to a great extent !**

# [B] Classification Accuracies

Article A				
Correct	Human	BOW	TF-IDF	BERT
<i>Synthetic</i>	<b>Real</b>	<b>Synthetic</b>	<b>Synthetic</b>	<b>Synthetic</b>

Greenland's government has decided not to apply for permission for further oil drilling in the coming year. This is announced by the Greenlandic Ministry of Nature, Environment and Agriculture in a press release. "We have decided not to apply for oil drilling in 2023, because we want to spend time developing a new strategy for the Greenlandic economy, which will form the basis for a new oil and gas strategy," it says. The government also emphasizes that it will maintain its "vision of a fossil-free Greenland". The decision comes after a meeting on Tuesday between the government's four parties. It is mainly the consideration for the climate that has led the government to drop further oil drilling.

17 out of 20 human participants classified as **Real**

# 05 Limitations

## DESIGN CHOICES FAVORING MACHINE DETECTORS

1. **Zero temperature sampling** → oversamples high-likelihood tokens
2. **Humans received *no* training**, classifiers were trained on +1000 labelled examples
3. **Humans saw real headings and subheadings** → evoked familiarity?

## GENERALIZABILITY OF THE STUDY

1. **Shortened articles** → not comparable to a real-world context.
2. **Useless generations?** Inferring factual information from just a headline requires additional context in reality.

# 05 Main Takeaways

## OUR STUDY SHOWS that ...

GPT-3 **can** be fine-tuned to produce Danish synthetic news articles **that are virtually indistinguishable** to real news articles for humans.

**BUT ...** the human eye is not all-seeing!

Constructing a machine detector for the same task, reveals that **machine detection of GPT-3 was possible to a great extent**

Likely related to underlying flaws in GPT-3's article generations

Different design could make results more favorable for humans

# Questions?



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