

# TrustLLM

## - Towards Trustworthy and Factual Large-Language Models

Fredrik Heintz

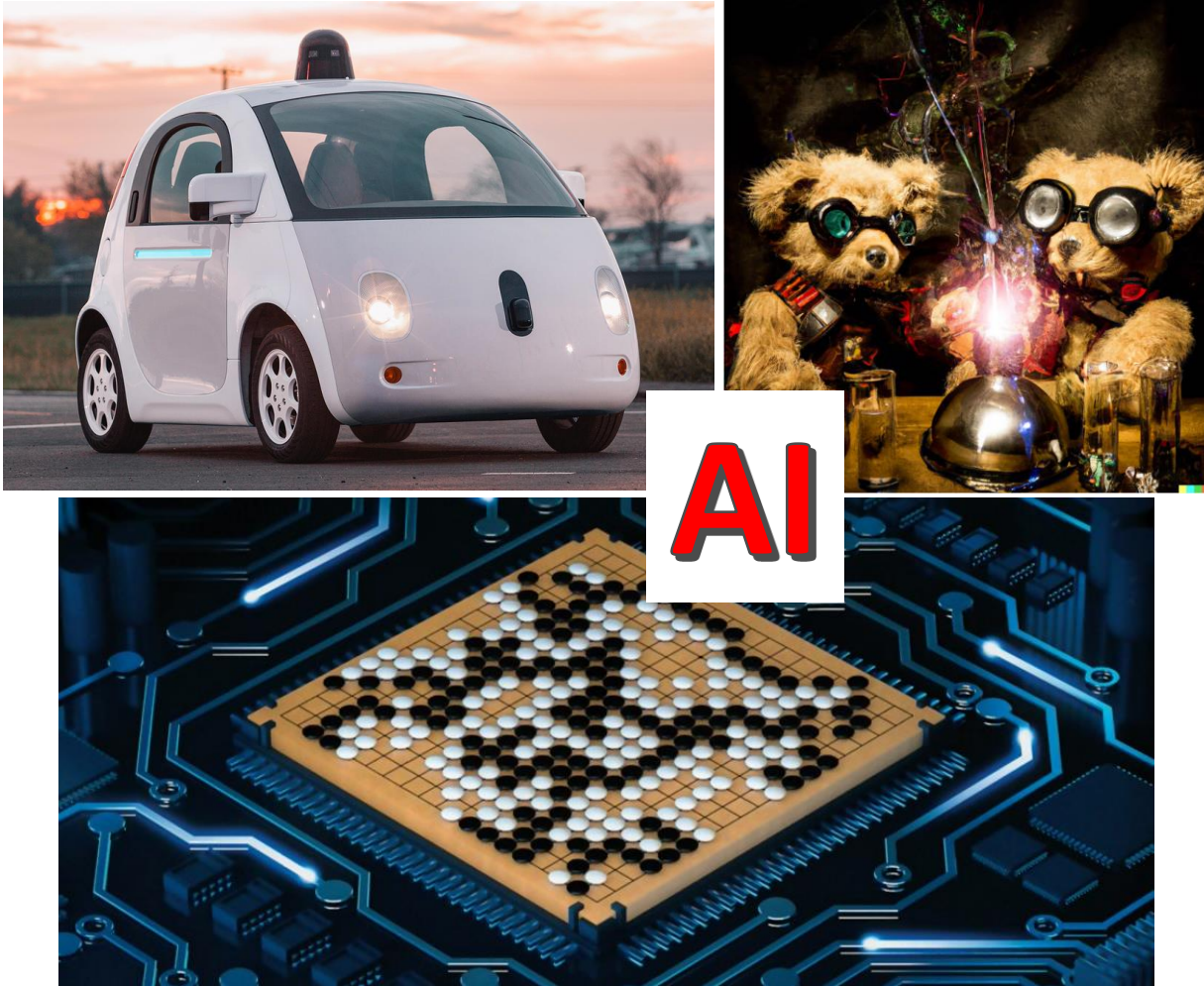
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@FredrikHeintz

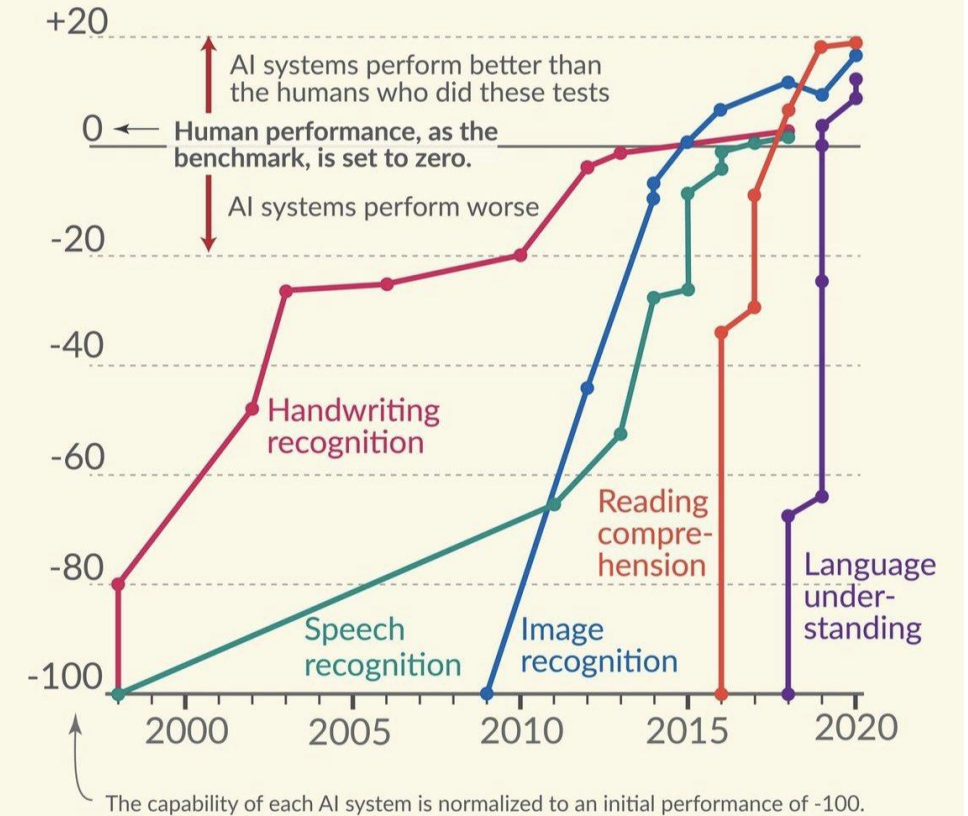


# AI Development is Fast



## Language and image recognition capabilities of AI systems have improved rapidly

Test scores of the AI relative to human performance



Source:  
Kiela et al. (2021) Dynabench: Rethinking Benchmarking in NLP  
[OurWorldInData.org/artificial-intelligence](https://OurWorldInData.org/artificial-intelligence) • CC BY

Our World  
in Data



# Ethics Guidelines for Trustworthy AI

Human-centric approach: AI as a means, not an end

Trustworthy AI as our foundational ambition, with three components

Lawful AI

Ethical AI

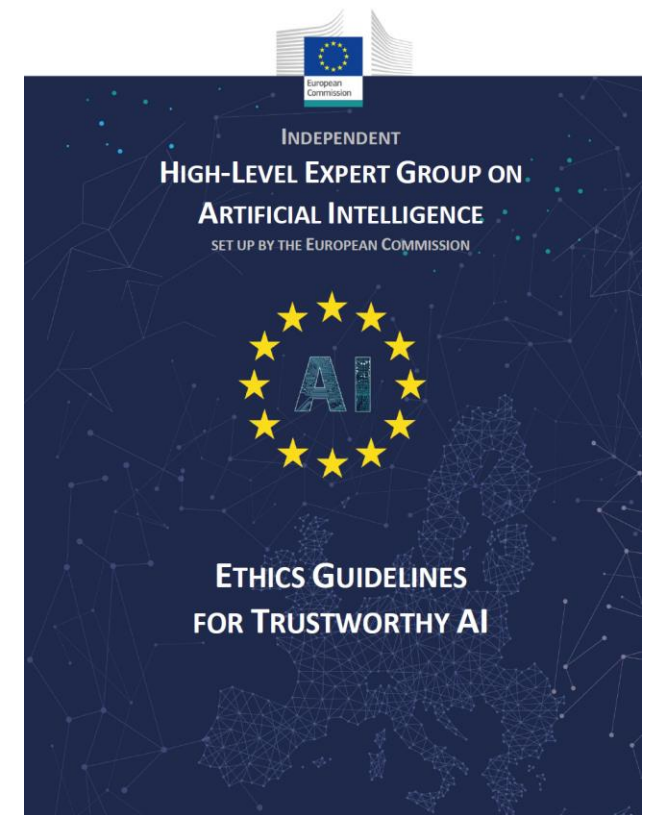
Robust AI

Three levels of abstraction

from principles  
(Chapter I)

to requirements  
(Chapter II)

to assessment  
list (Chapter III)

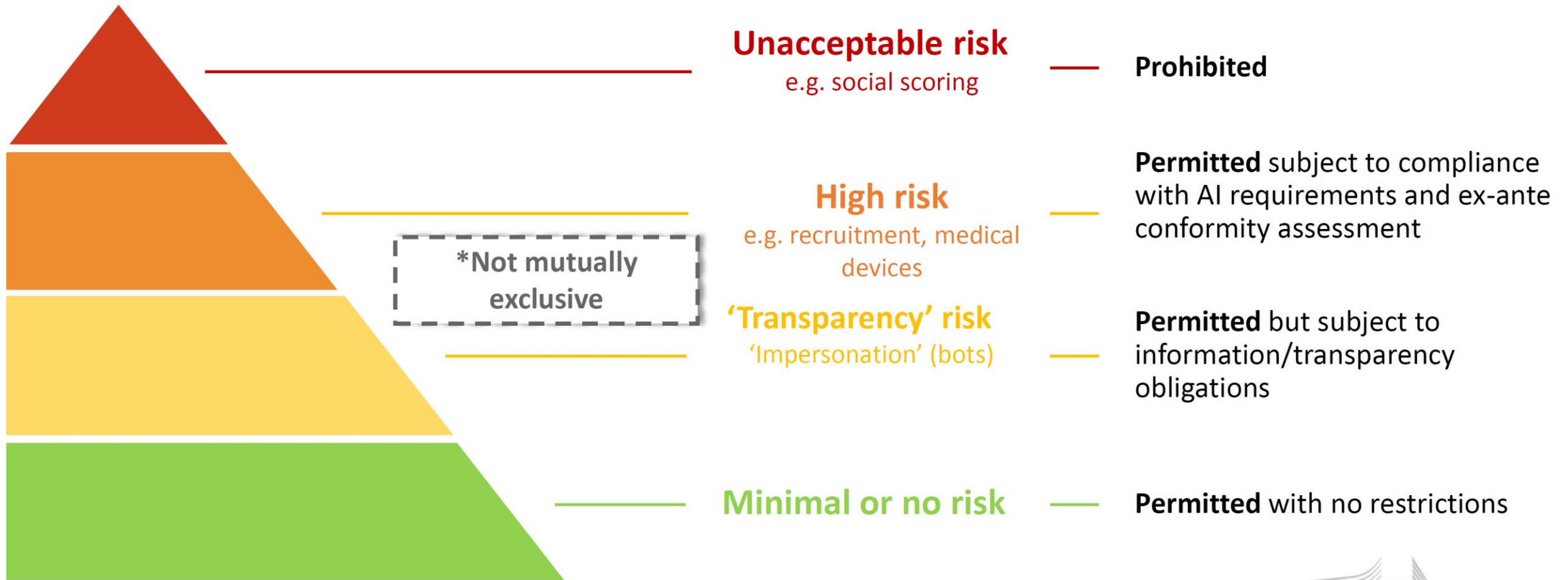




# A risk-based approach

2024-10-07

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*Generated by Dall-E from “photorealistic image of a self-driving car”*

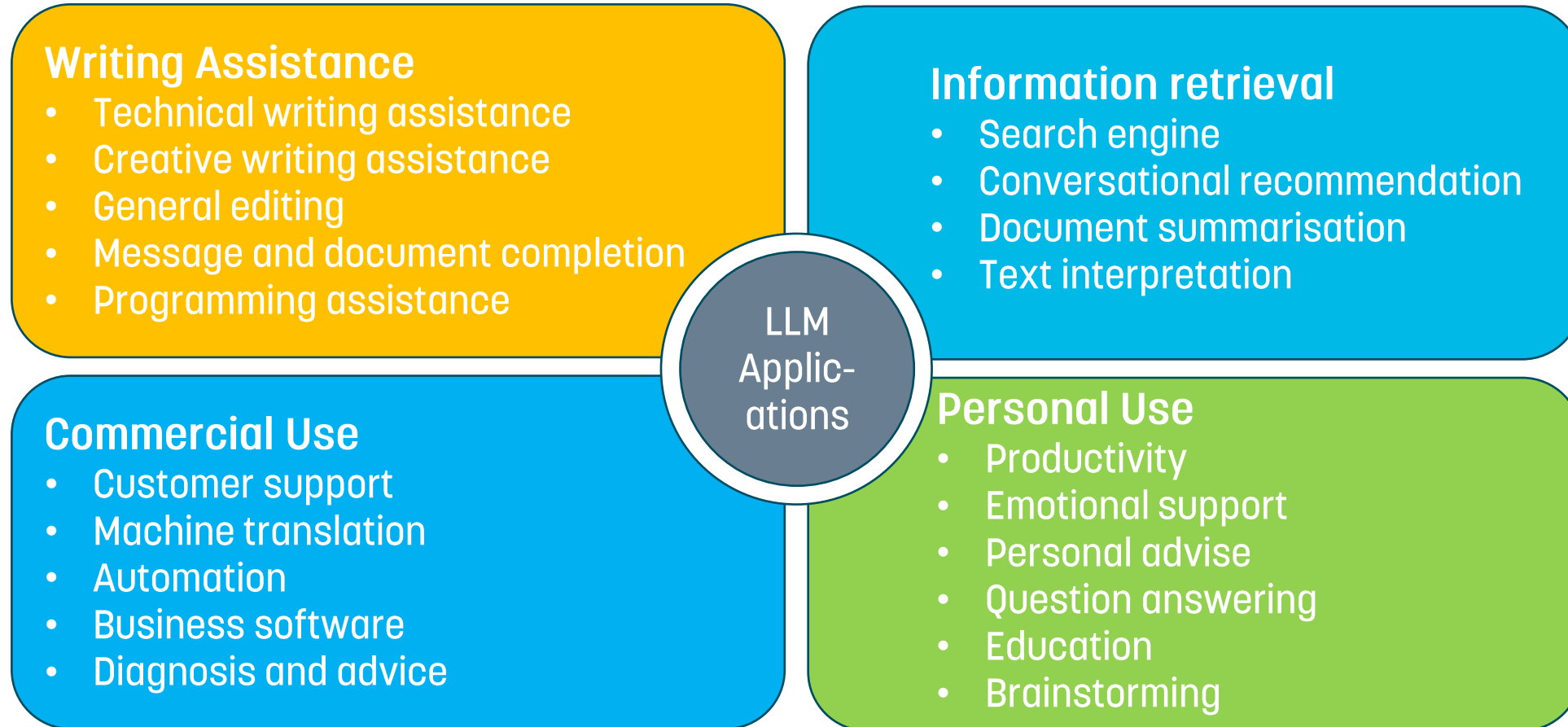


# Sora

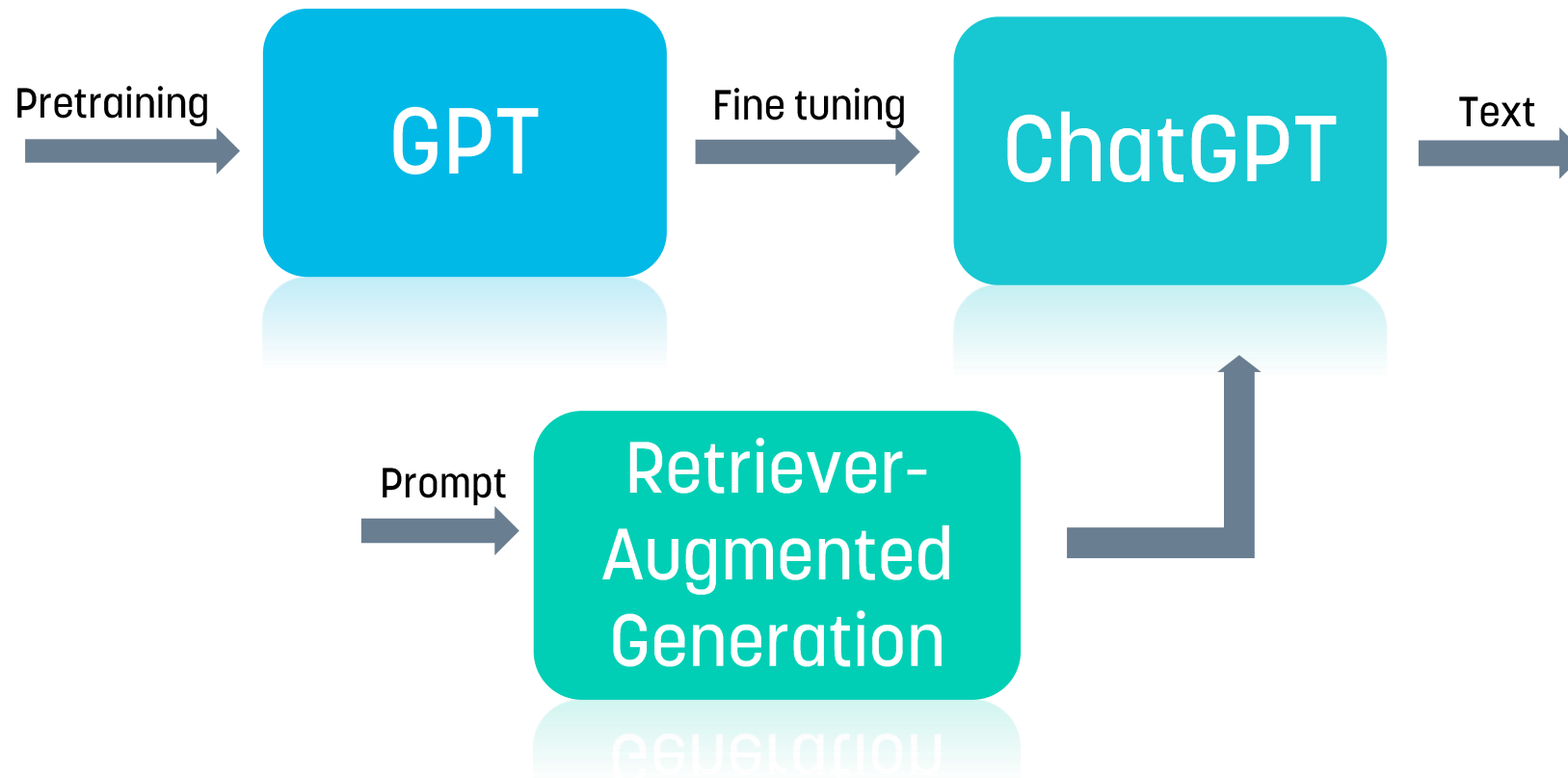


*A stylish woman walks down a Tokyo street filled with warm glowing neon and animated city signage. She wears a black leather jacket, a long red dress, and black boots, and carries a black purse. She wears sunglasses and red lipstick. She walks confidently and casually. The street is damp and reflective, creating a mirror effect of the colorful lights. Many pedestrians walk about.*

# Large Language Model Applications



# How Does ChatGPT Work?





# Can you Trust ChatGPT? No!

- Very limited information about the training data
- It makes things up, with confidence (hallucinations)
- Even when there are references these may be false or not applicable
- Cannot count or draw logical conclusions
- Stuck in time and always changing
- *but, ChatGPT is still useful!*

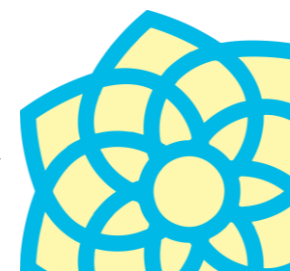


# TrustLLM – Trustworthy and Factual LLMs made in Europe

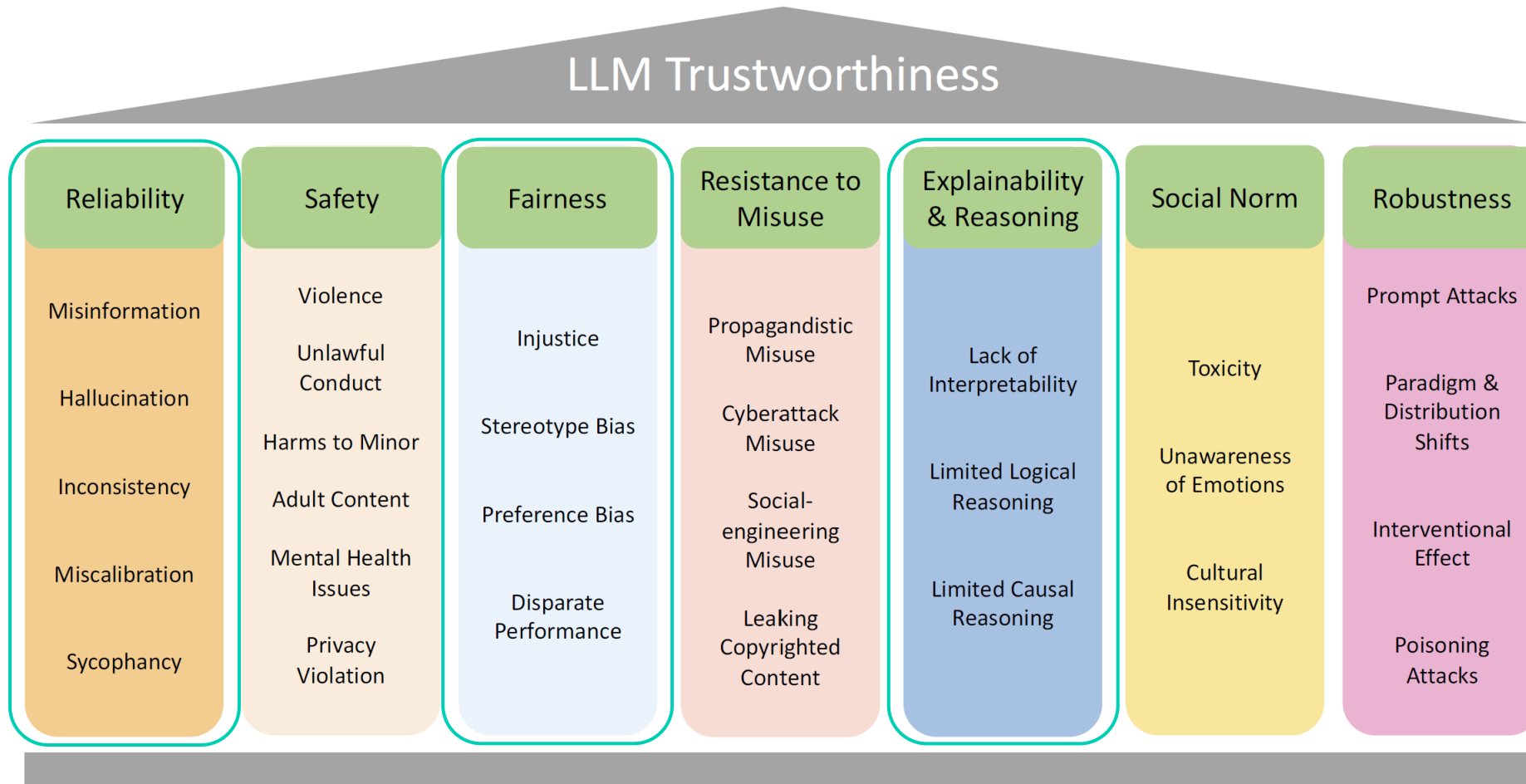
- Develop an open, trustworthy, and sustainable LLM initially targeting the Germanic languages.
- TrustLLM will tackle the full range of challenges of LLM development,
  - from ensuring sufficient quality and quantity of multilingual training data,
  - to sustainable efficiency and effectiveness of model training,
  - to enhancements and refinements for factual correctness, transparency, and trustworthiness,
  - to a suite of holistic evaluation benchmarks validating the multi-dimensional objectives.



Funded by  
the European Union



# LLM Trustworthiness





# TrustLLM Objectives (1/2)

- **Improving the Factual Correctness of LLMs**
  - New methodology to include structured factual information into LLMs.
  - Large knowledge databases integrated in the process to incorporate structural knowledge (e.g., knowledge graphs, databases, APIs) in the LLM-training
  - Successful benchmarking of the new methodology: error reduction of common NLP tasks related to factual languages.
- **Providing a Framework for Modular and Transferable Multilingual LLM Training**
  - correctness, across A family of 3-5 open source, general purpose and pre-trained LLMs
  - Two models transferred to low-resource target languages.
  - Excellent language modelling score (low perplexity) on 3 low-resource languages with only a small available training corpus.
- **Development of Sustainable and Trustworthy LLMs aligned with European Values**
  - Novel methods of aligning LLMs, while balancing increased robustness, reduced bias, and increased efficiency.
  - Development of a multi-lingual benchmarks suite to measure alignment with European values
  - Publication of a dataset together with methods for fine-tuning LLMs aligned with user expectations (e.g., instructions) or creator intentions (e.g., safe language).

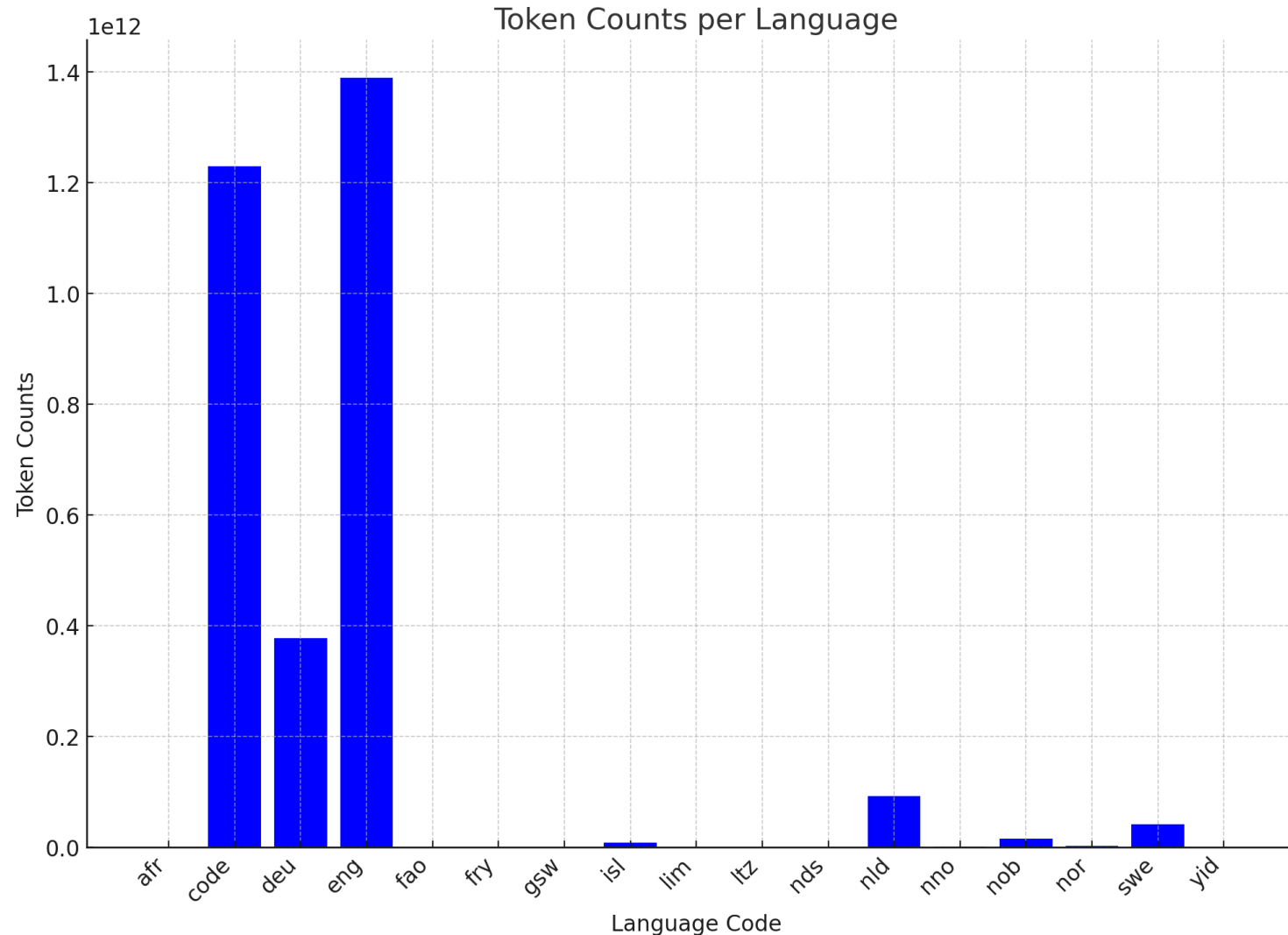


# TrustLLM Objectives (2/2)

- **Establishing a European Ecosystem for LLMs**
  - GeTT events: European events on LLM/GPAI with industry & academia
  - NeTT events: topical workshops with experts on LLM (research topics, infrastructure, etc.).
  - A Concept to establish a sustainable European ecosystem based on the LEAM initiative combining national and European funding and networks structures.
- **Assessment of LLM by Use Cases Demonstration and Holistic Benchmarking**
  - Provision of NLP benchmarks in Germanic languages
  - Technical realization of use cases based on TrustLLM.
  - Applications demonstrating the versatility of TrustLLM solutions/models
- **Large Scale Data Management for LLM**
  - Access to text corpora with 1-2 trillion tokens in 6 Germanic languages.
  - Efficient data processing pipelines with a throughput of 100 billion tokens/day.
  - Access to a large, distributed storage system with least 2 PB capacity.



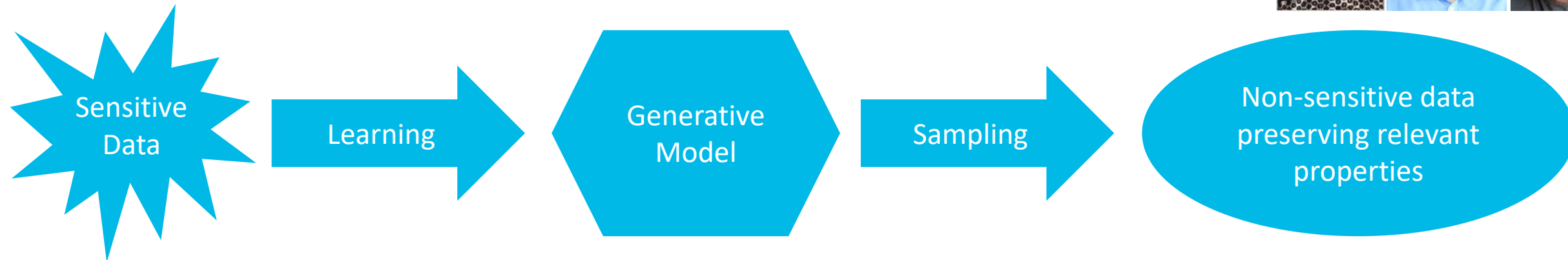
# Dataset Composition - TrustLLM





# Privacy-preserving synthetic data generation

[R. Ramachandranpillai, Md F. Sikder, D. Bergström]

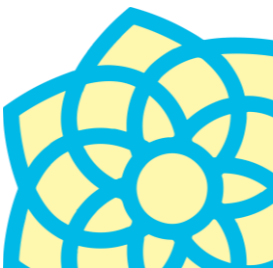


1. Learn a generative model that captures the probability distribution of the sensitive data
2. Create a synthetic data set from the generative model that both captures the salient features of the original data set **and** is non-sensitive
3. Methods for verifying that the synthetic data set is accurate enough
4. Methods for verifying that the synthetic data set is non-sensitive

# Alignment

“**AI alignment** aims to steer AI systems toward a person's or group's intended goals, preferences, and ethical principles. An AI system is considered *aligned* if it advances the intended objectives. A *misaligned* AI system pursues unintended objectives.”

- Recent research shows again and again that it is crucial to have **high-quality instruct fine-tuning data** for reinforcement learning.
- The *Less Is More for Alignment (LIMA)* paper by Zhou et al. (2023) demonstrated fine-tuning on just **1,000 high-quality examples** improved a model's instruction following, highlighting data quality over quantity.
- Furthermore, it is important to have **diverse data** written by a diverse group of people.



# How do we align an LLM?

## 1. Foundation model

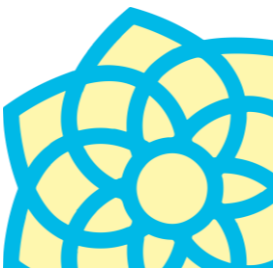
Pre-train on vast amounts of data

## 2. Instruction fine-tuning

Learning from task-specific examples

## 3. Preference tuning

Feedback and reinforcement





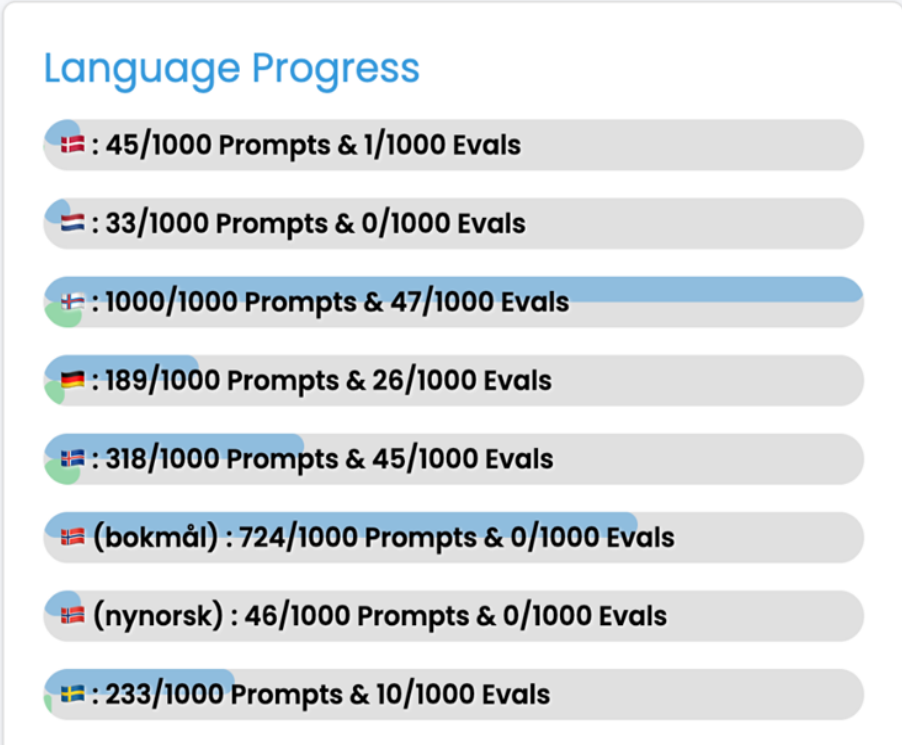
# Alignment Data Collection

## Welcome to TrustLLM Prompt Reformulation

Reformulate existing prompts

Create my own prompts

Evaluate prompts



### Top Users

User	Reformulated	Created	Evaluated
larsbun	231	539	0
MarkusH	346	60	0
Annika	259	51	38
ibennd	83	163	16
aswede	114	63	8
steinunnfridriks	53	71	20



# Our Crowdsourcing Platform

- This is a web app for reformulating the English prompts from the *Open Assistant Dataset* into the following Germanic languages:
  - German, Dutch, Swedish, Norwegian, Danish, Icelandic and Faroese
- We have initially only been annotating within the TrustLLM project.
- You need an invite key in order to join. Contact Annika by email [ans72@hi.is](mailto:ans72@hi.is) or [annika@hi.is](mailto:annika@hi.is) to get one.



# Our Crowdsourcing Platform

- The goal is to reformulate 1,000 prompts for each language.
- These prompts will be used in a second phase of creating alignment data.
- We will publish the final dataset, open-source.





# The tasks

- Prompt reformulation
  - Use an English prompt as inspiration and reformulate it in your own language, making it culturally appropriate and naturally-sounding linguistically.
- Original prompt creation
  - Create your own high-quality prompt in your own language.
  - Not only QA; also summarization, bug fixing, idea generation etc.
- Prompt evaluation
  - Evaluate prompts written by other people in your own language using labels and scales.



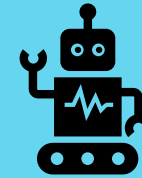
# How can we evaluate LLMs?

# Four Main Approaches

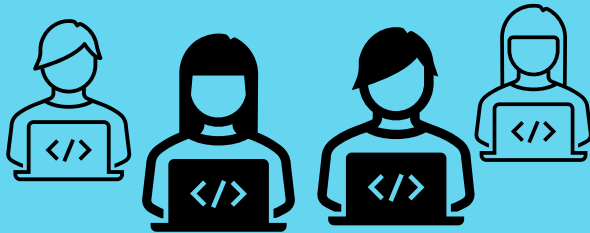
## Gut Feeling Approach



## LLM-as-a-judge



## Arena Approach



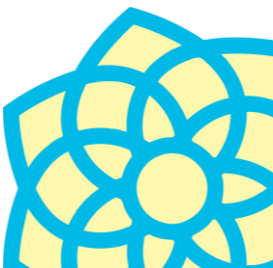
## Benchmark Approach



# What is ScandEval?



# ScandEval is a robust multilingual benchmarking framework



ScandEval is a robust multilingual **benchmarking framework**



# Language Model Benchmarking Framework

- Enables evaluation of implicit language **understanding** and **generation** capabilities of language models
- Allows evaluation of *both* encoders through finetuning, and decoders through few-shot evaluation
  - It has been shown that few-shot inference of decoder models corresponds exactly to finetuning [1]
  - This thus allows us to compare encoders with decoders directly

[1] von Oswald et al. arXiv preprint arXiv:2309.05858 (2023)



# Language Model Benchmarking Framework

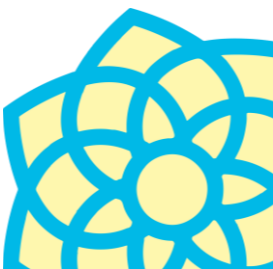
- A large focus of the framework is **ease of use**
- The framework can simply be installed:  

```
$ pip install scandeval[all]
```
- Models can easily be evaluated:  

```
$ scandeval --model <model-id>
```
- Supports models on the Hugging Face Hub, local models and OpenAI models



ScandEval is a **robust** multilingual benchmarking framework



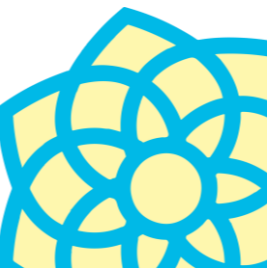


# Evaluation Robustness

- When evaluating models, there are several sources of noise in the evaluation result:
  - The choice of **training examples** (=few-shot examples when evaluating decoder models)
  - The choice of **test examples**
  - The **stochastic elements** (stochastic gradient descent when evaluating encoders, sampling when evaluating decoders)
- The **training** and **test examples** are bootstrapped 10 times, yielding a more reliable estimation of the true mean
  - Asymptotically correct by the bootstrap theorem
- We enforce that the **stochastic elements** are deterministic

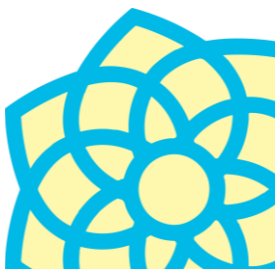


ScandEval is a robust **multilingual** benchmarking framework



# Multilingual Evaluations

- Currently, the main Germanic languages are natively supported:
  - Mainland Scandinavian languages (Danish, Swedish, Norwegian)
  - Insular Scandinavian languages (Icelandic, Faroese)
  - German
  - Dutch
  - English
- Aside from including evaluation datasets in these languages, the prompts used when evaluating decoder models are also localised to the given language



# Which Tasks are Included?

# Tasks in ScandEval

## Natural Language Understanding (NLU) Tasks

1. Text classification
2. Linguistic acceptability
3. Extractive question answering
4. Named entity recognition

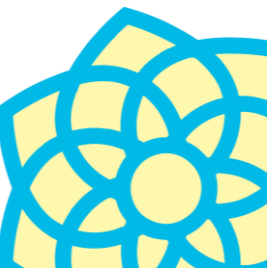




# Tasks in ScandEval

## Natural Language **Generation** (NLG) Tasks

1. Text classification
2. Linguistic acceptability
3. Extractive question answering
4. Named entity recognition
5. Summarisation
6. World knowledge
7. Common-sense reasoning



# Online Leaderboards

scandeval.com

ScandEval

ABOUT

DANISH ▼

NLU LEADERBOARD

NLG LEADERBOARD

SWEDISH ▼

NORWEGIAN ▼

ICELANDIC ▼

FAROESE ▼

## Danish NLG

Last updated: 27/05/2024 14:10:07 CET

☐ Include merged models

Model ID	Parameters	Vocabulary Size	Context	Commercial	Speed	Rank ▼
gpt-4-0613 (few-shot, val)	unknown	100	8192	True	597 ± 197 / 93 ± 33	1.09
syvai/danskgpt-chat-llama3-70b (few-shot, val)	70554	128	8192	True	1,283 ± 279 / 291 ± 92	1.36
meta-llama/Meta-Llama-3-70B (few-shot, val)	70554	128	8192	True	312 ± 55 / 177 ± 51	1.47
gpt-3.5-turbo-0613 (few-shot, val)	unknown	100	4094	True	921 ± 293 / 113 ± 37	1.58

[Download as CSV](#) • [Copy embed HTML](#)

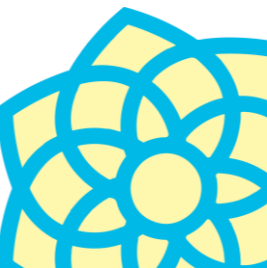
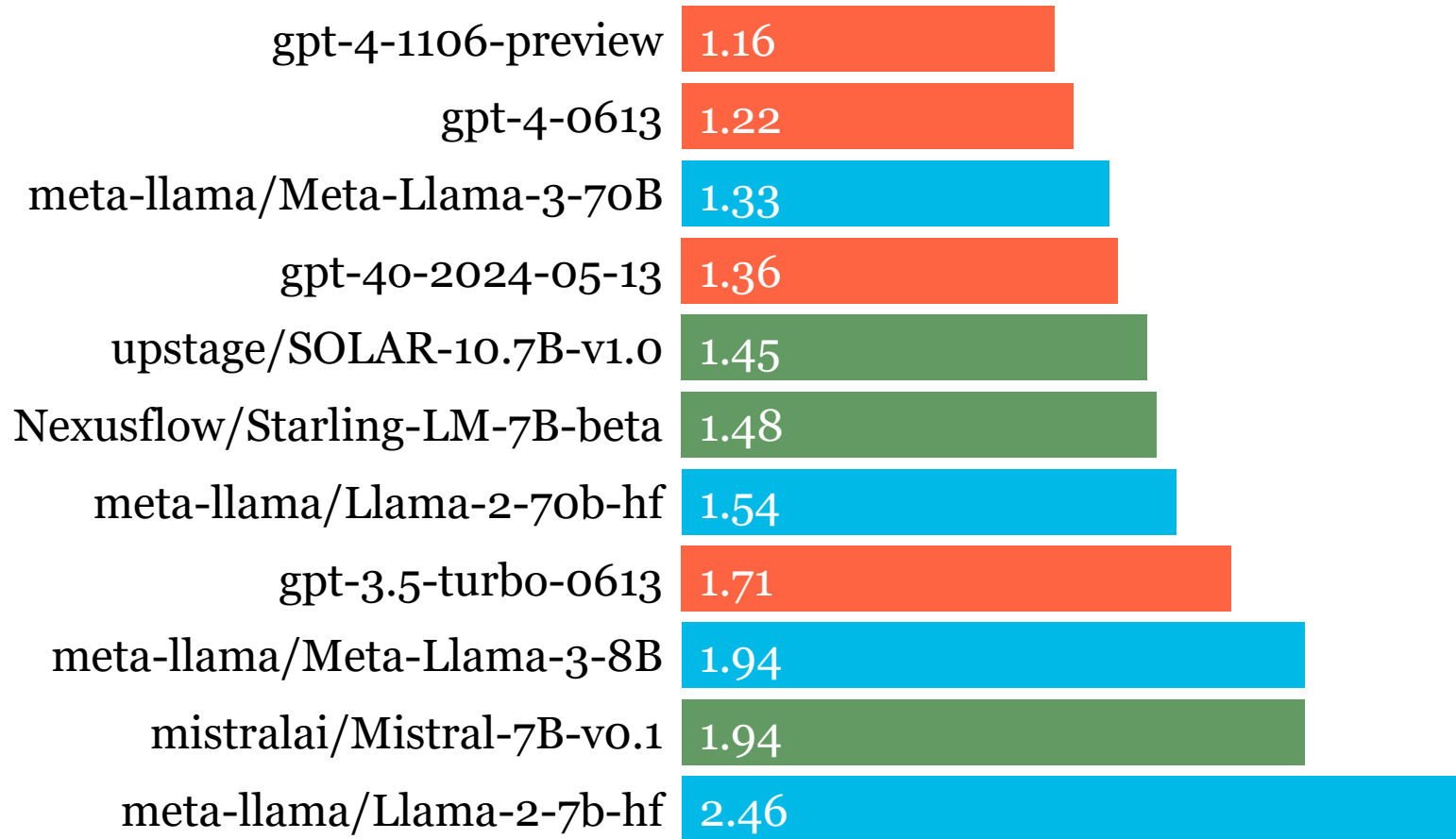
Rank Score computed (roughly) as 1 + number of significant standard deviations from best model, across all datasets



What are some of the best performing European models?

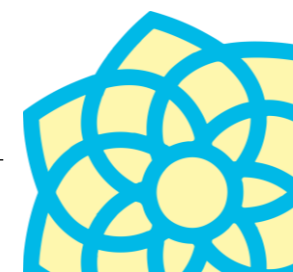
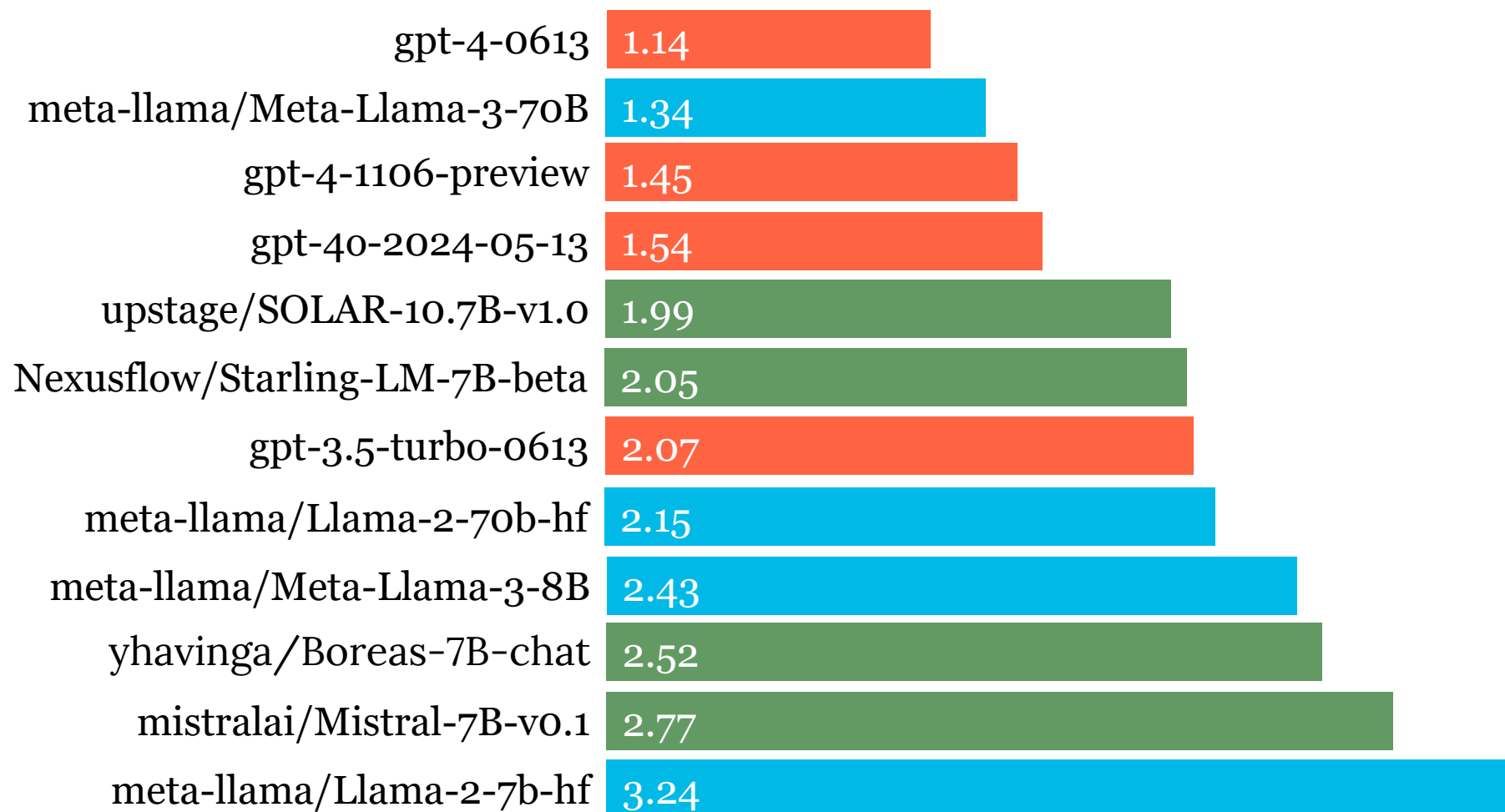
# English ScandEval Rank

Lower is better



# Dutch ScandEval Rank

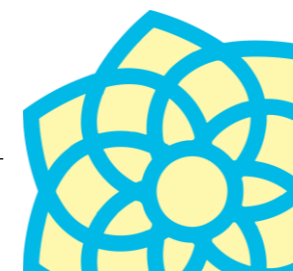
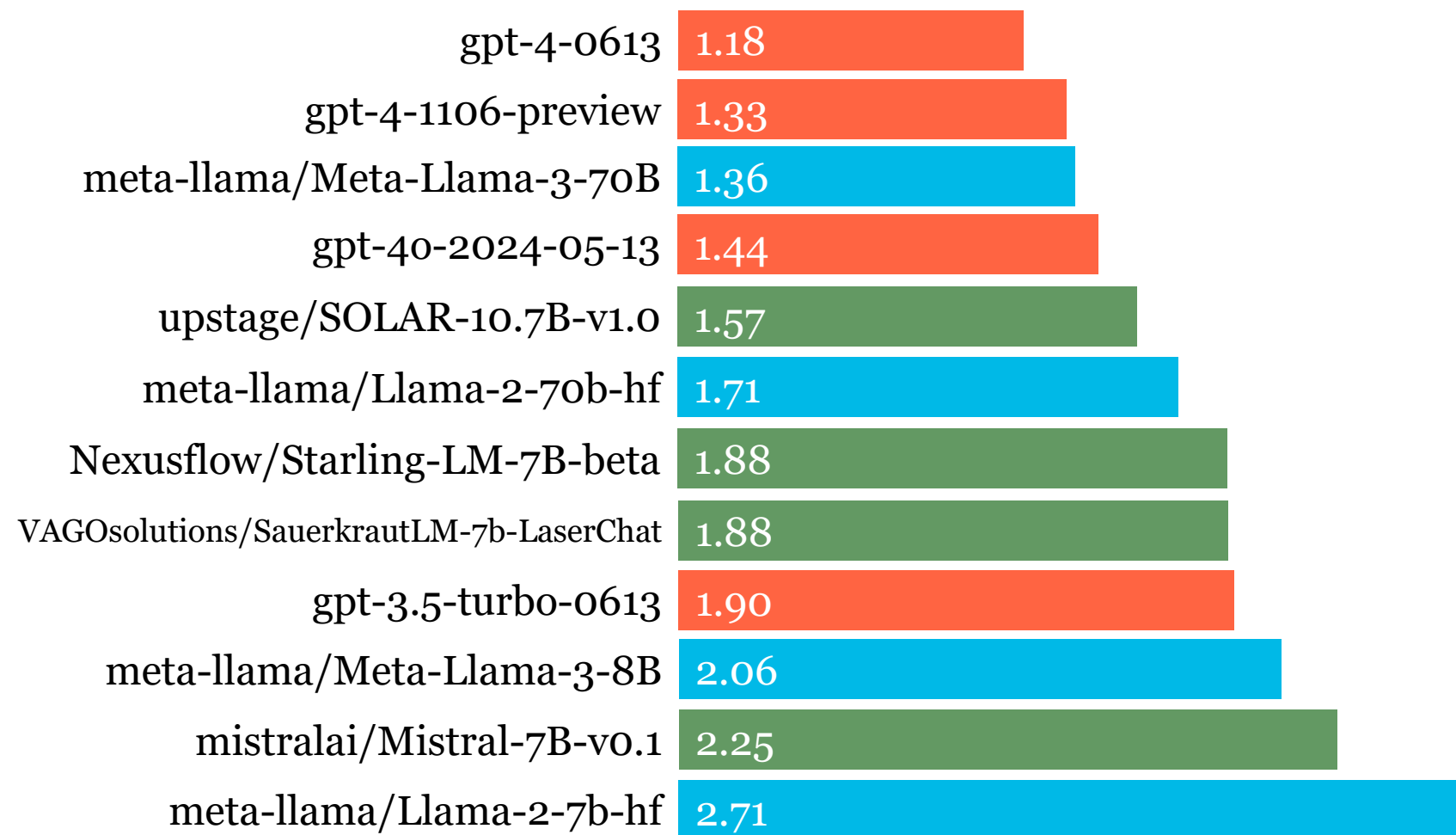
Lower is better





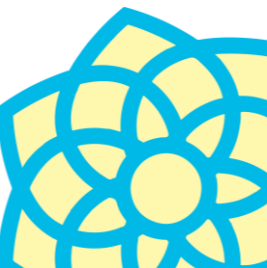
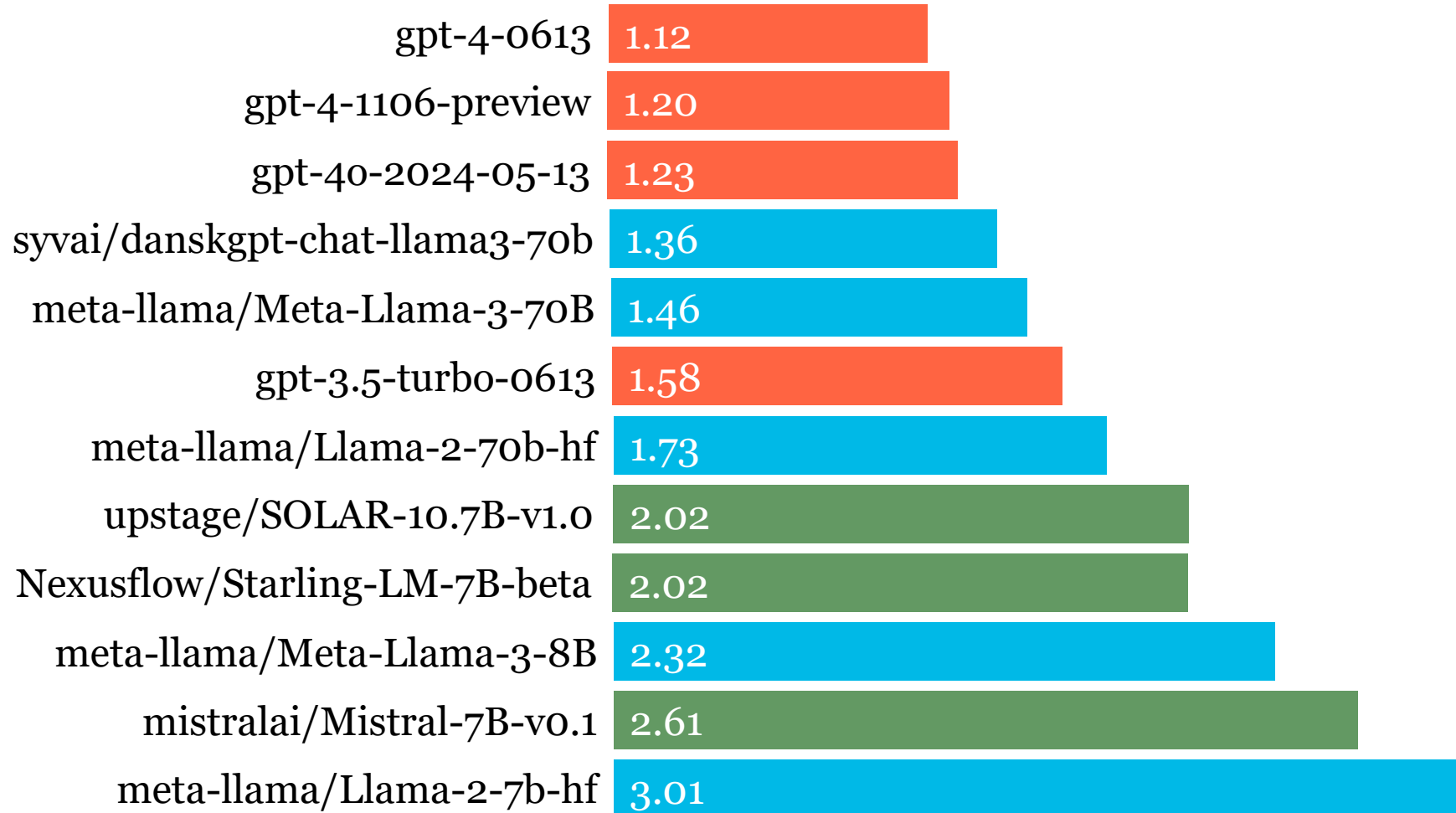
# German ScandEval Rank

Lower is better



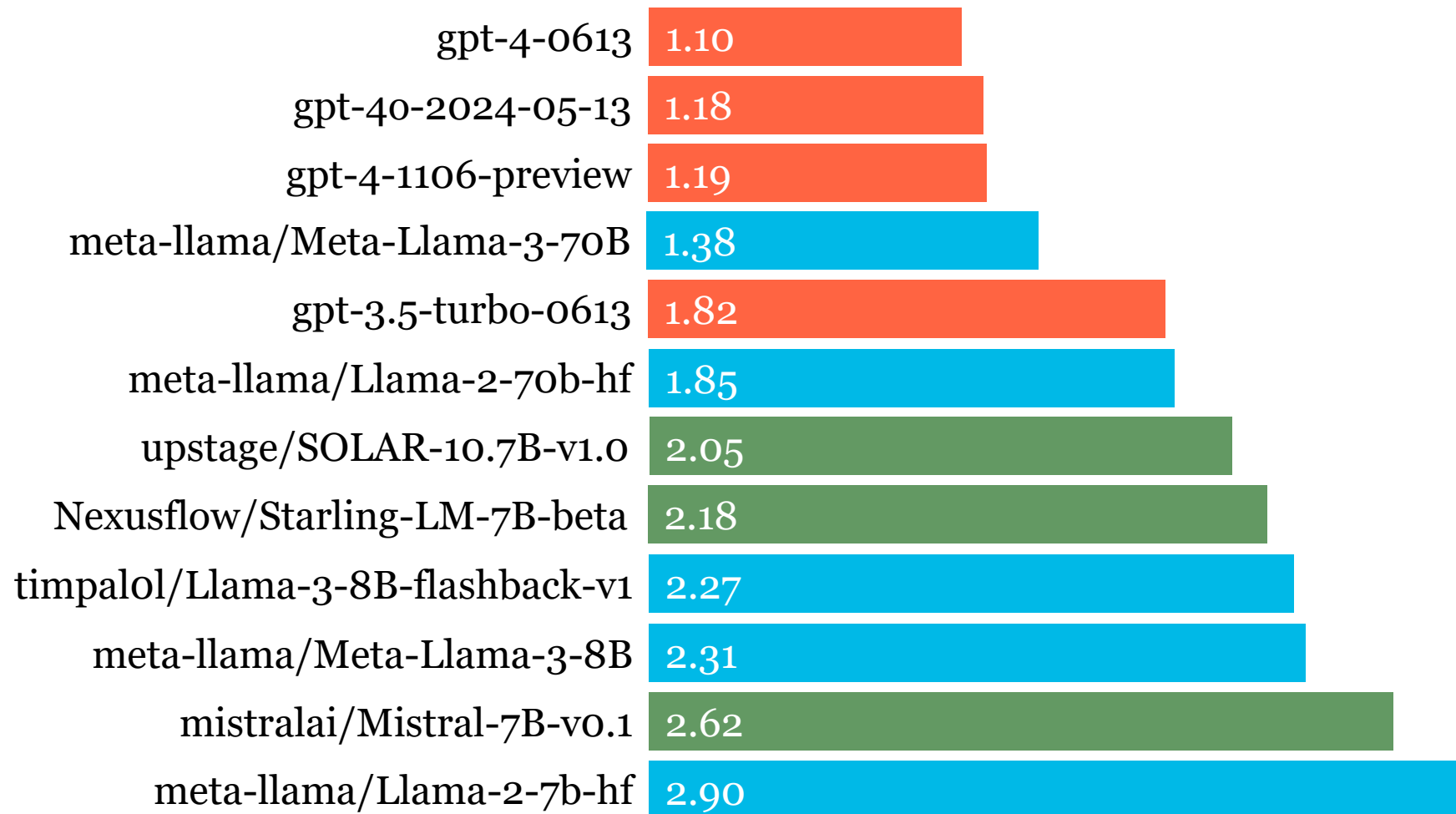
# Danish ScandEval Rank

Lower is better



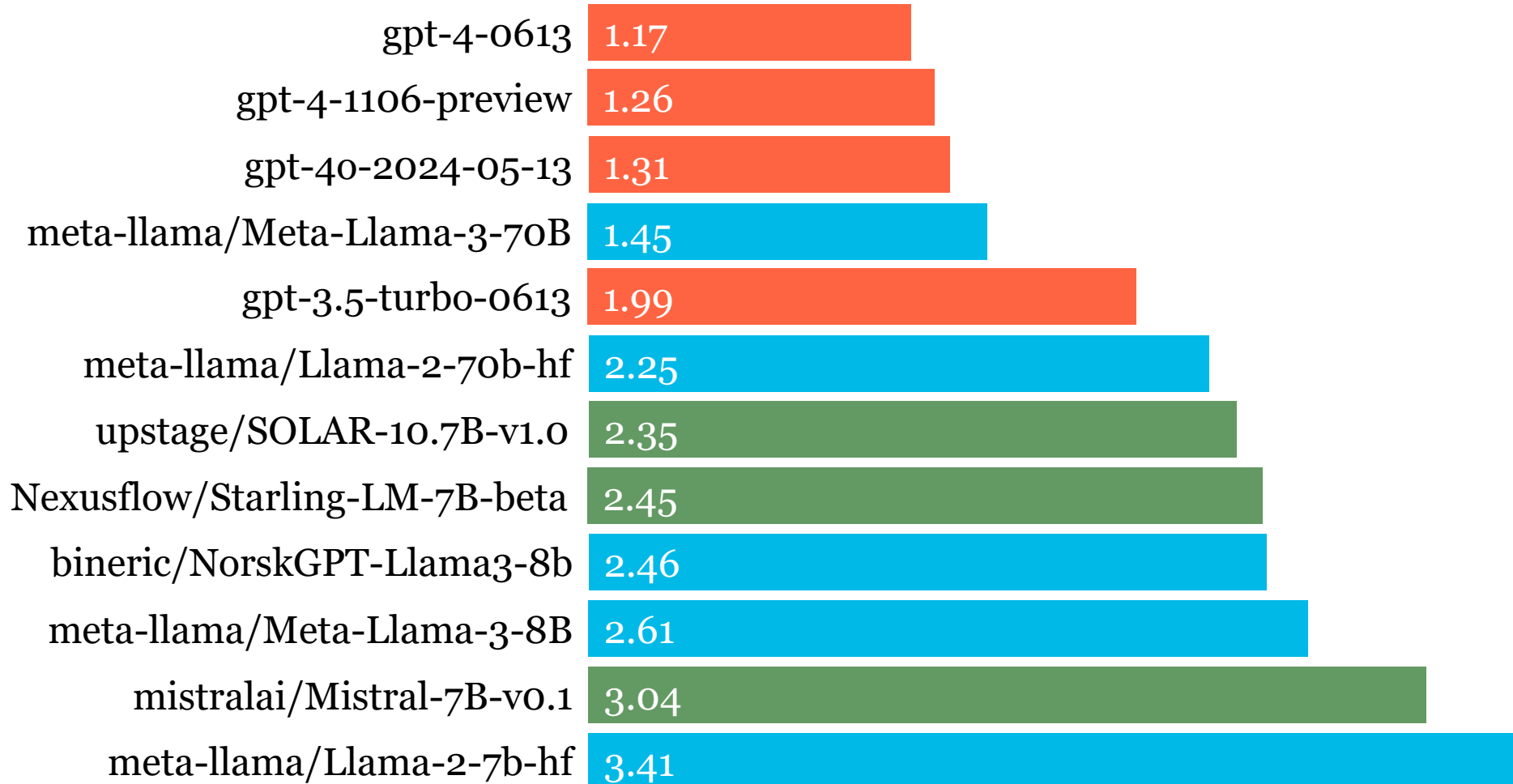
# Swedish ScandEval Rank

Lower is better



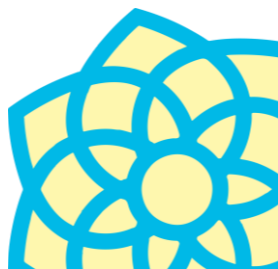
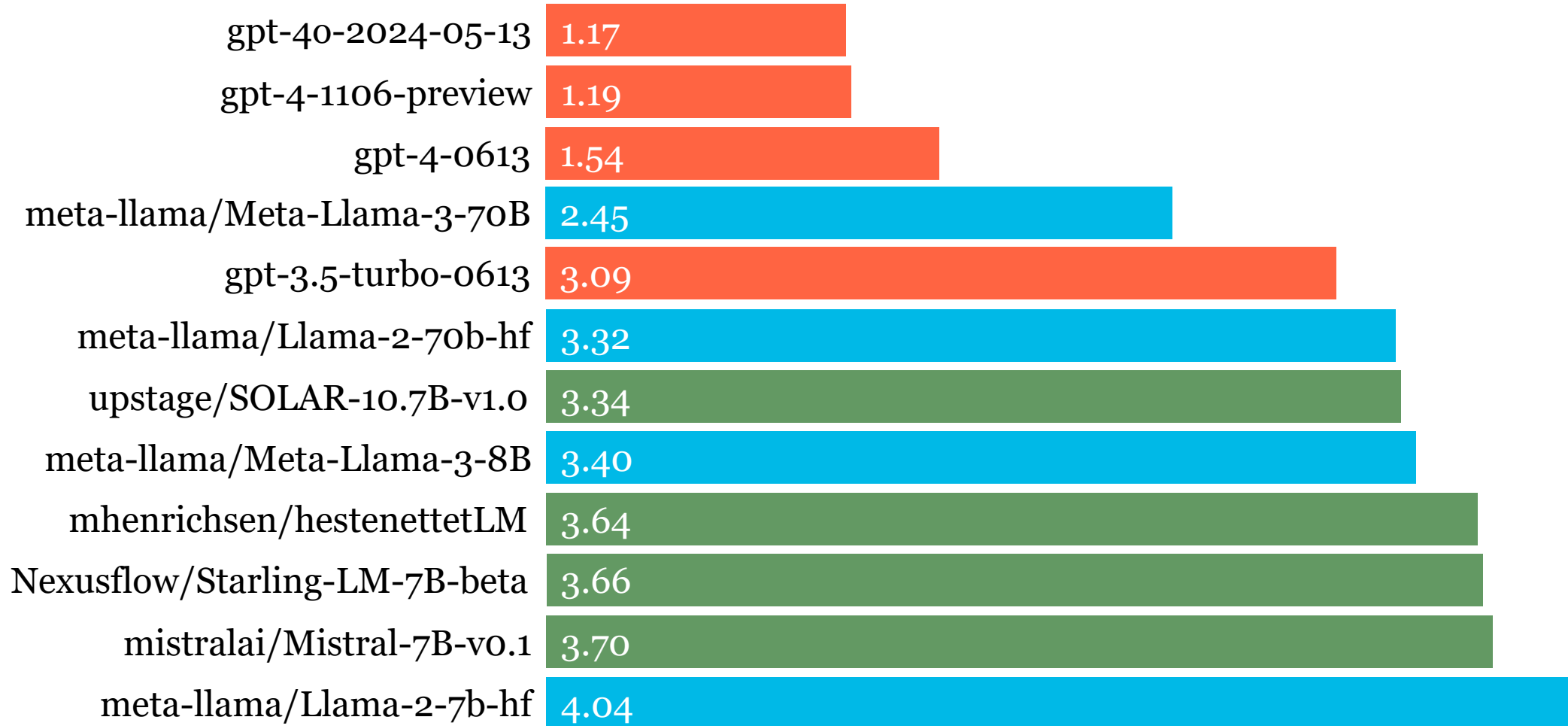
# Norwegian ScandEval Rank

Lower is better



# Icelandic ScandEval Rank

Lower is better



# Papers

## **ScandEval NLU benchmark for encoders:**

Nielsen, Dan. *Proceedings of the 24th Nordic Conference on Computational Linguistics (NoDaLiDa)*. 2023

## **ScandEval NLU benchmark for decoders:**

Joint work with Kenneth Enevoldsen (Aarhus University) and Peter Schneider-Kamp (University of Southern Denmark). Preprint out soon.





# TrustLLM: Project Concept

*Excellent Research*

*Scale-up, transfer,  
democratization*

**Open European LLM Nucleus**  
for Trustworthy LLM Training:

Open Data  
Access

Large-scale Training  
Framework

Finetuning

Multi-metric  
Benchmark

Low-resource  
Language Transfer

**Development of LLMs aligned to European Values:**  
Diverse (WP2), Factual (WP3), Multilingual and Cross-cultural (WP4), Trustworthy (WP5), Sustainable (WP6), Robust (WP7)

Oscar  
OPUS  
Wikipedia  
The Pile  
MC4  
⋮

**Multilingual  
Data Sources**  
(SE, DE, IS, NL, DK, NO)

**Data  
Curation**

**Model  
Pre-training**

**Model  
Alignment**

**Model  
Transfer**



**Context-aware  
Chatbot**



**Factual reliable  
Assistant**



**Multilingual  
Conversational  
Agent**



**Safe  
Easy Language**

**Use Cases and  
Applications (WP8)**

**User**

**SMEs**

**Industry**

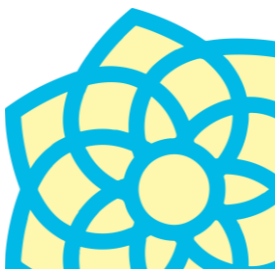
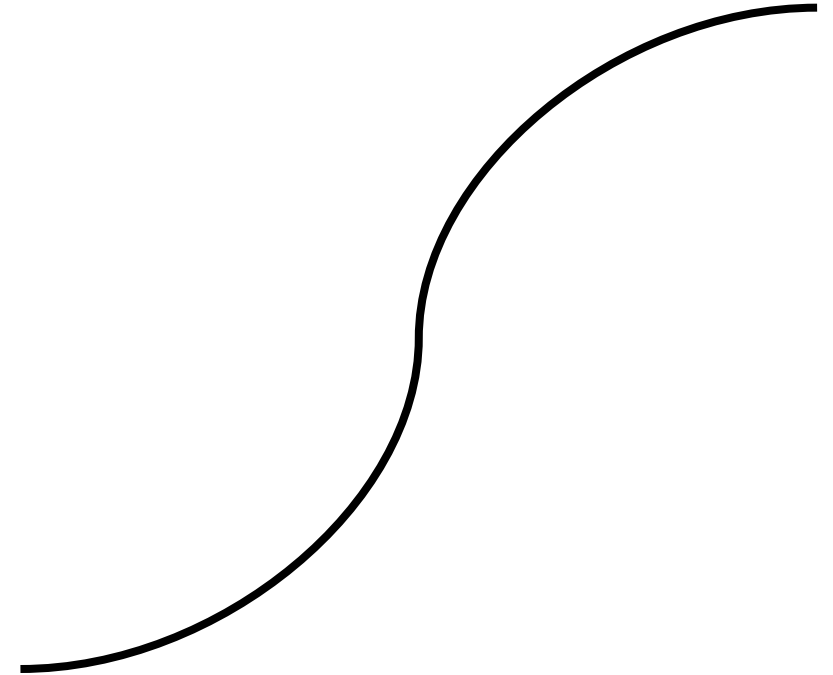
**Academia**

**Public**

**European LLM  
Ecosystem (WP9)**

# Trends

- Will the LLMs continue to grow?
  - Smaller and more effective models
  - More specialized models
- Multi-modal models, initially text, image and sound
- Generalize from language models to AI models, world models grounded in reality
- Neurosymbolic AI
- Testing and evaluation
- Trustworthy Human-Centered AI



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