AI for subject indexing and descriptive cataloguing – a civilized approach?

Osma Suominen

KI – konkurrent eller kompanjong?
Oslo, 24 April 2026

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About me

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Osma Suominen

Information Systems Specialist, National Library of Finland

Doctoral thesis "*Methods for Building Semantic Portals*"
Semantic Computing Research Group, Aalto University, 2013
Supervisor Professor Eero Hyvönen

finto

Joined the National Library in 2013
to set up the Finto.fi thesaurus and ontology service

Working on opening up bibiliographic metadata as Linked Data (Fennica-LD) and automated subject indexing (Annif)

Open source software projects e.g.:

- Skosify – Validation and QA tool for SKOS vocabularies
- Skosmos – SKOS vocabulary publishing tool
- Annif – Tool for automated subject indexing and classification

Languages spoken: Finnish, Swedish (-Norwegian), English, Estonian

OUR VISION: **Bildung** at the heart of society

Swedish: **Bildning** (dannelse) i kärnan av samhället

Photo: NatLibFi / Frida Lönnroos

VS.

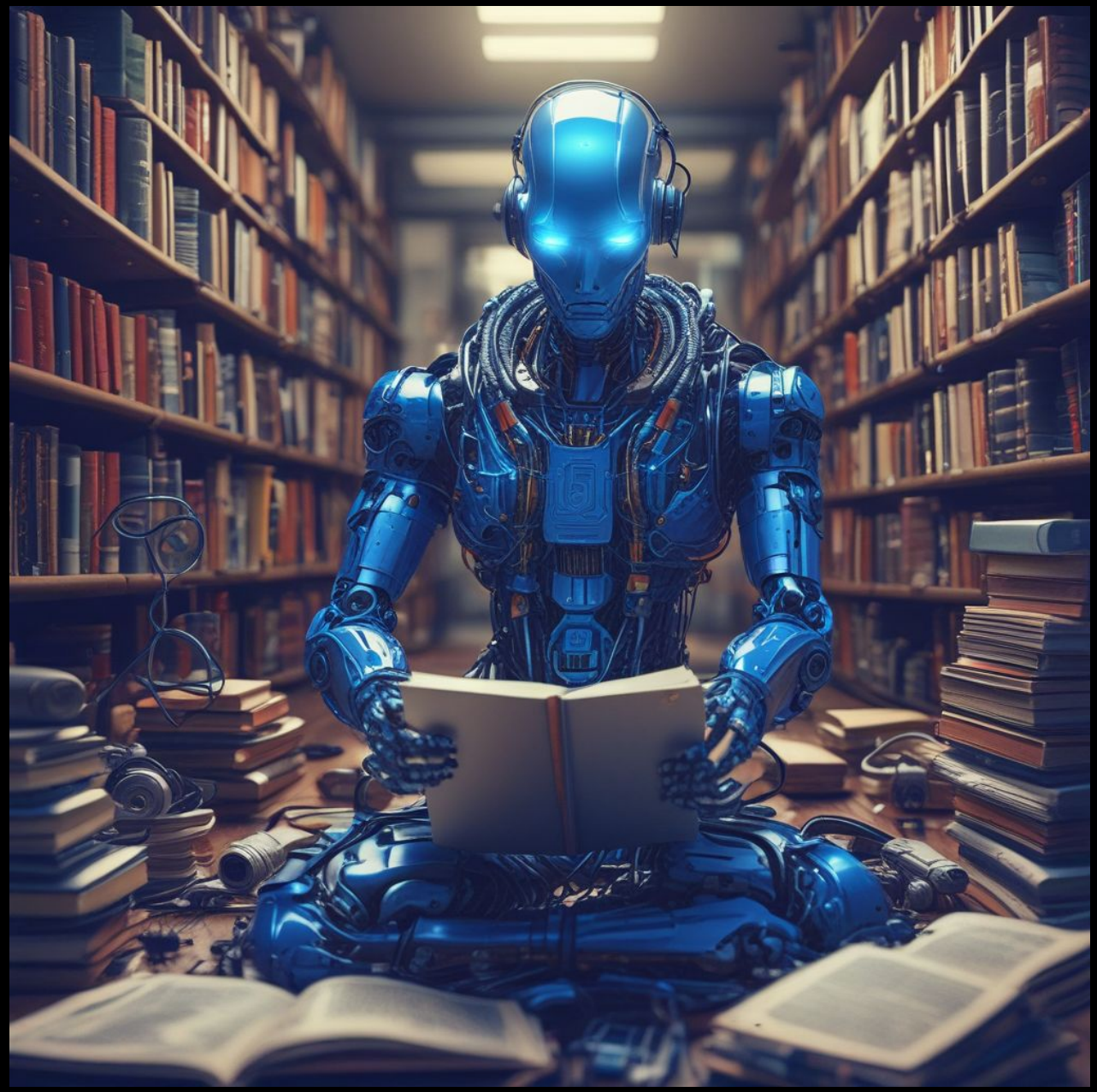
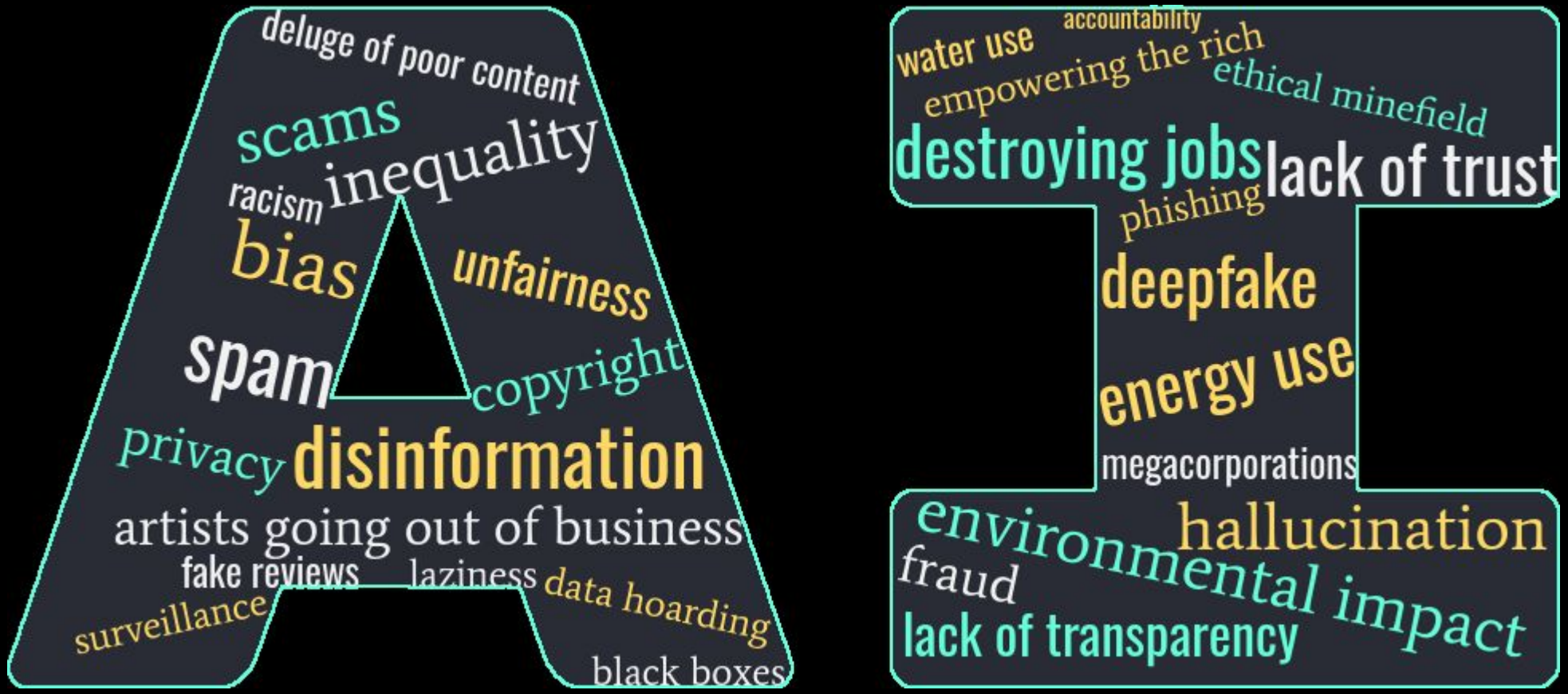


Image: Stable Diffusion XL
Prompt: An AI robot looking at books, electric blue, mechanic, shiny, lots of cables

There are many problems with AI...



As a 🧑‍💻👩‍🔧, I am ✨🤩 about AI

As a 👩‍🏫🤔, I am 😱🥶 about AI

How can we develop practical AI solutions without amplifying the worst problems of AI?

Automated subject indexing: Annif and Finto AI

How can I annotate my documents or data with concepts from Finto?

annif

developed since 2017

General purpose open source **tool** for automated subject indexing and classification

Multilingual, supports many vocabularies

Code on GitHub, website with test form and API

Global development and user community; user forum **annif-users** on Google Groups

annif.org

fintoai

launched in 2020

Automated subject indexing **service** for production use, based on Annif

Supports indexing with the General Finnish Ontology YSO in Finnish, Swedish and English language

Web user interface and API service

Intended to support subject cataloguers in Finland regardless of institution (GLAMs, public administration); sister project to the Finto vocabulary service

ai.finto.fi

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Demonstration

Automated subject indexing using Finto AI

Based on the Annif toolkit that we develop

Languages: Finnish, Swedish or English

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No LLM needed! *

* In fact, LLMs are pretty bad at subject indexing when prompted in a typical naïve way.

We took part in the two LLMs4Subjects challenges in 2025.
Our hybrid system (Annif + data pre-processing with LLMs) performed better than pure LLM systems.

Metadata extraction and descriptive cataloguing: Meteor + LLM and BIBRA

How can I get basic metadata out of electronic grey literature publications for making a catalogue record?

Grey literature?

reports

working papers

government documents

white papers

preprints

theses

...

Semi-formal non-commercial PDFs published on the web – lots of them! Often with lacking metadata.

Image made with DreamStudio (based on Stable Diffusion)
Prompt: "A big pile of papers, reports, documents, PDF files, word documents, powerpoint slides, posters, articles and books forming a wave in the style of Hokusai"

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Why extract metadata from grey literature PDFs?

1. help users of **digital repositories** who need to enter metadata when uploading PDFs

2. in **web archiving**, get more information about harvested PDF files

3. ease **cataloguing** of e.g. government reports and doctoral theses

Extracting bibliographic metadata

PDFs

system

title: Novel methods for earth fault passage indication in non-effectively grounded electricity distribution networks

creator: Farughian, Amir

publisher: University of Vaasa

faculty: School of Technology and Innovations

date: 2022

e-issn: 2323-9123

p-issn: 0355-2667

e-isbn: 978-952-395-055-9

p-isbn: 978-952-395-054-2

ispartofseries: Acta Wasaensia

numberinseries: 500

KANSALLISKIRJASTO
NATIONALBIBLIOTEKET

Our solution

1. FinGreyLit data set for training and evaluation
   currently 1,600 PDFs with curated metadata, 4 languages, on GitHub

2. GreyLitLM series of fine-tuned small language models
   fine-tuned using 75% of the FinGreyLit documents

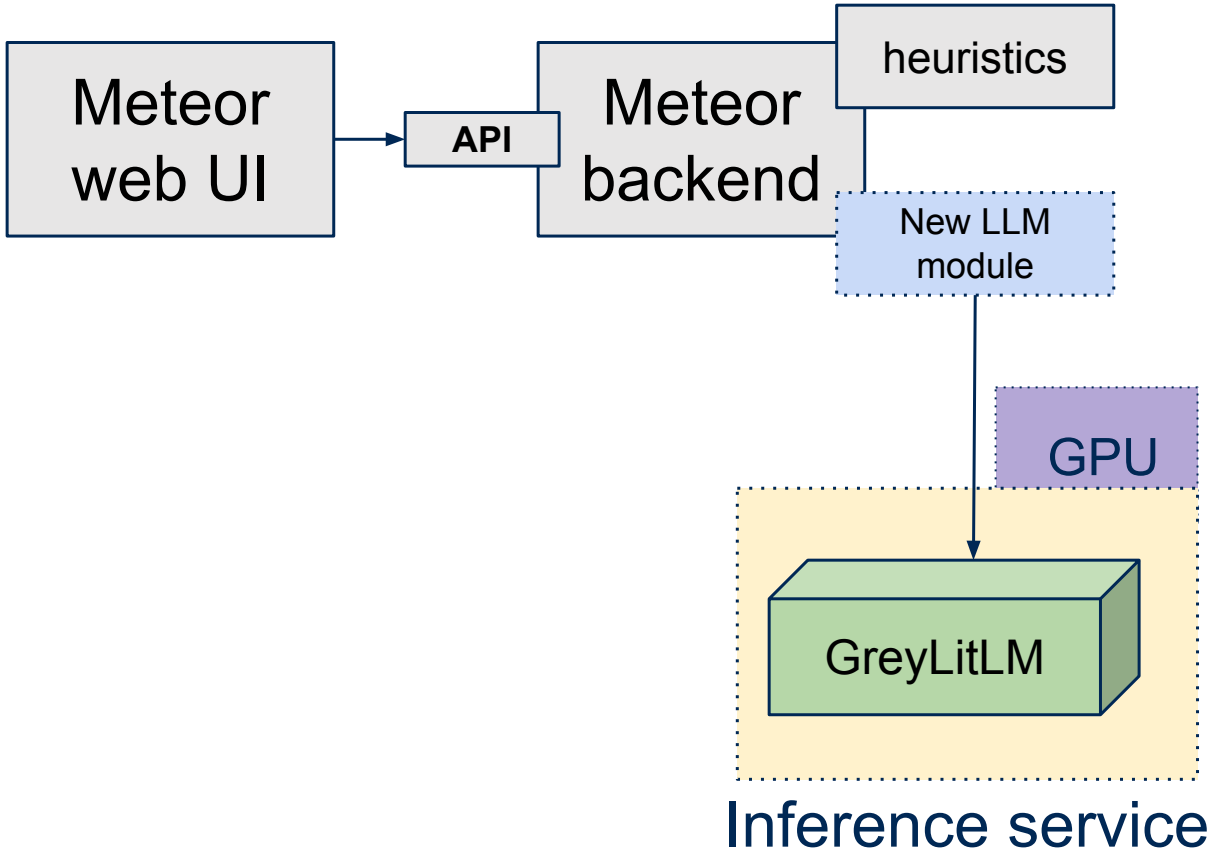
3. Meteor tool, extended to use an external LLM service

Meteor metadata extraction tool

- published on GitHub as Open Source by the National Library of Norway
- extracts title, authors, language, publisher, year, ISBN, ISSN
- main target is government / public sector reports in Norwegian and English
- based on hardcoded heuristics, no machine learning
- simple web UI + REST API for integration, used in production at NLN

Extending Meteor with an **external LLM**

- Meteor built-in heuristics didn't work well on Finnish grey literature
- We extended it to use an external LLM service with GreyLitLM



Demonstration

Extraction of metadata from PDF publications part 1

Prototype system

Small fine-tuned language models (here based on Gemma3-4B)

Extension of the Meteor tool by the National Library of Norway

Evaluation – does it work?

Extraction methods to compare:

1. Empty baseline – predict nothing; effective for seldom used fields, e.g. DOI.
2. Meteor – hardcoded logic and heuristics in the Meteor tool.
3. Qwen2.5 0.5B – extremely small language model, fine-tuned by us.
4. Gemma3 4B – very small language model, fine-tuned by us.
5. Mistral NeMo 12B – small language model, fine-tuned by us.

Main metric: **Average per-field F1 score across test set documents (n=400)**

Separate overall averages calculated for:

1. 7 fields extracted by Meteor tool
2. 13 fields extracted by the fine-tuned language models

Table 1
Per-field evaluation results (average F1 scores) for different metadata extraction methods. We have calculated the averages separately for the seven fields supported by Meteor and all 13 fields extracted by language models.

	baseline	Meteor	Qwen2.5 0.5B	Gemma3 4B	Mistral NeMo 12B
language	0%	97%	99%	99%	**100%**
title	0%	40%	74%	87%	**92%**
alt_title	76%	-	83%	**87%**	**87%**
creator	17%	65%	79%	87%	**89%**
publisher	8%	8%	76%	80%	**86%**
year	14%	72%	89%	91%	**93%**
e-isbn	63%	83%	89%	**92%**	90%
e-issn	78%	86%	92%	92%	**94%**
p-isbn	72%	-	92%	**92%**	**92%**
p-issn	83%	-	96%	**96%**	**96%**
doi	91%	-	99%	**100%**	**100%**
type_coar	0%	-	80%	82%	**89%**
average (7 fields)	26%	65%	86%	90%	**92%**
average (13 fields)	42%	-	87%	90%	**92%**

Saturation point reached?
Larger LLMs won't help much

Beyond Meteor: BIBRA

Meteor provides a platform for quick experimentation, but we need to do more:

- User interface for metadata experts
- Support for basic extraction + turning the result into a bibliographic record (e.g. MARC, BIBFRAME, Dublin Core)
- REST API with the above functionality
- Modular backends (many different methods), as in Annif
- Test and evaluate methods using CLI, as in Annif

→ First step: BIBRA toolkit that we just started to develop

Demonstration

Extraction of metadata from PDF publications part 2

Prototype system

BIBRA tool, new user interface

Behind the scenes:
- REST API
- modular toolkit, to be extended

🤔 Is this civilized AI?

5 points

for building civilized AI

1. Use AI to make the world better

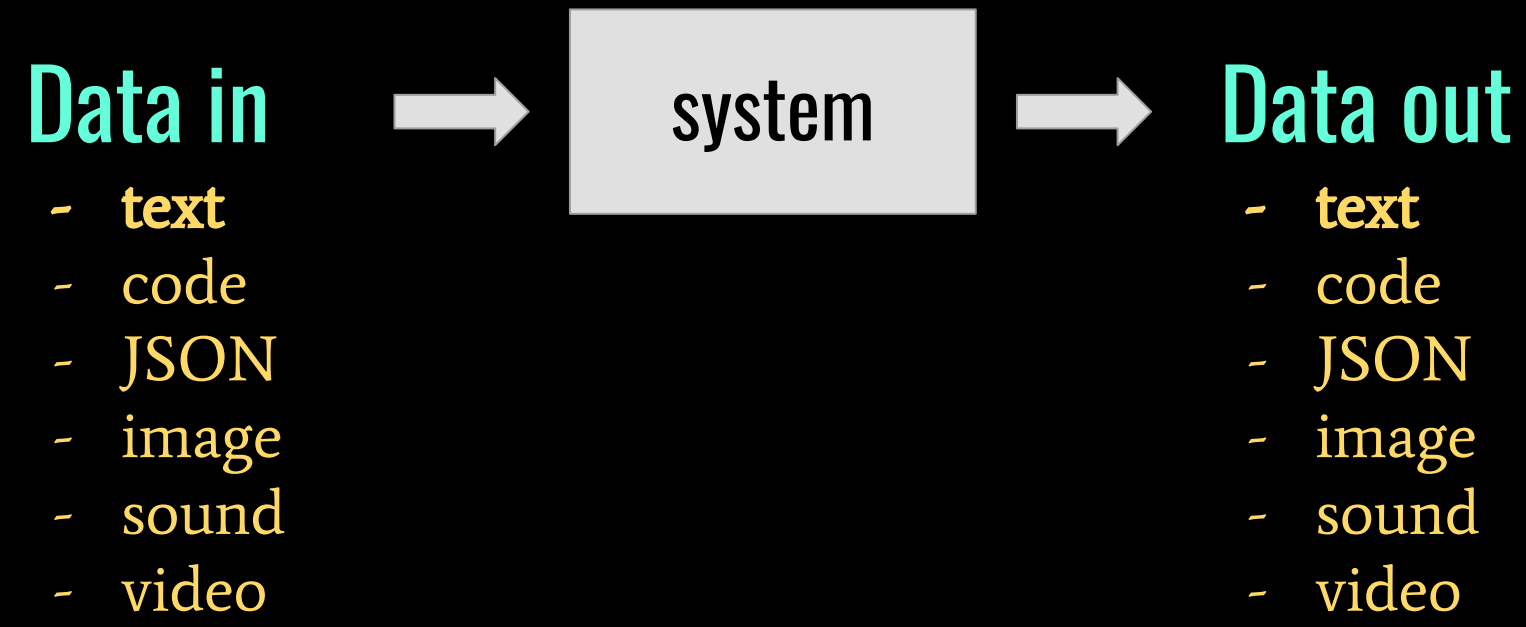
AI is not the goal

Human in the center

What positive changes can we make via AI?

2. Use the smallest AI that works

AIs can solve data-oriented challenges

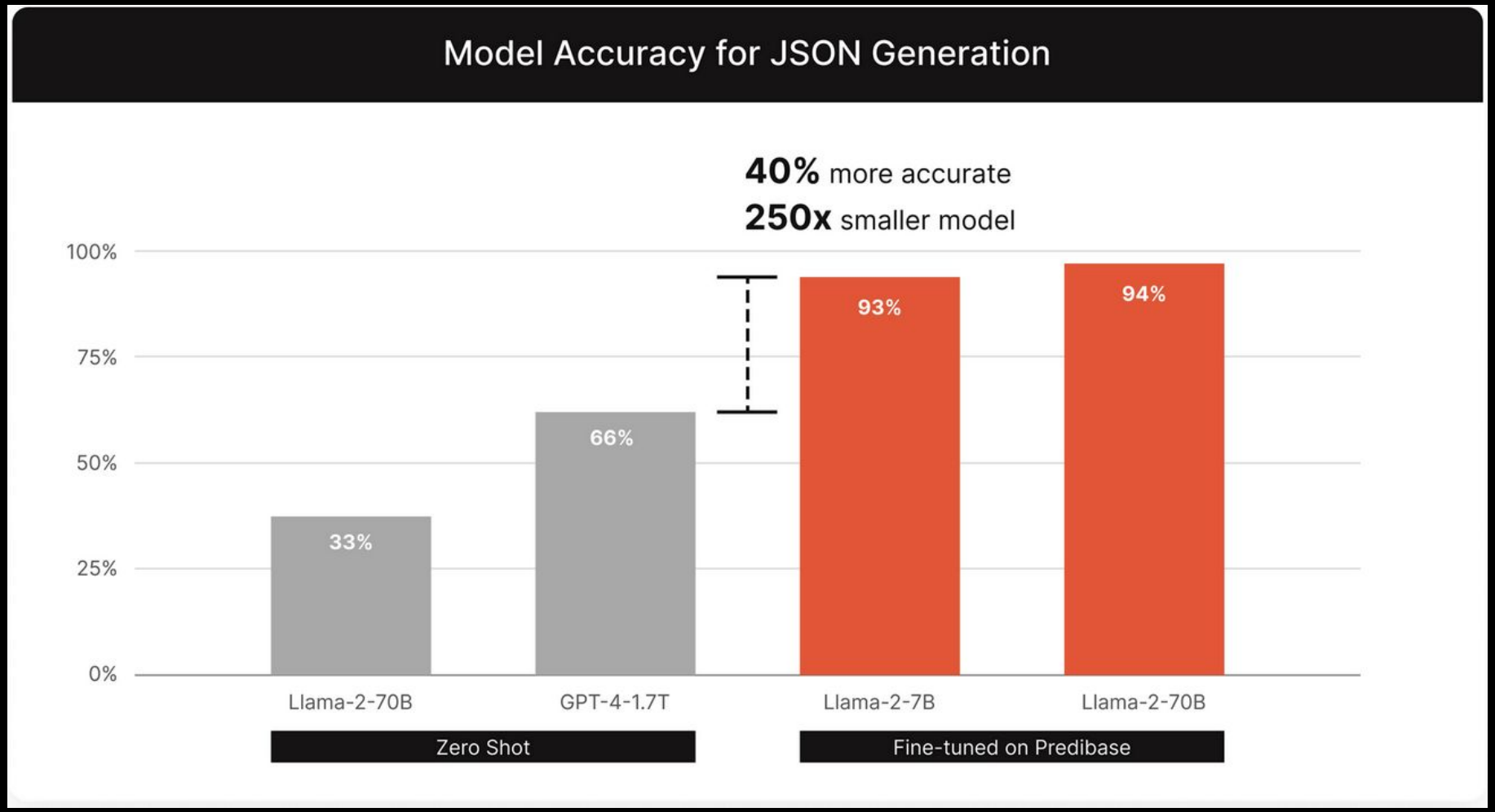


Try lighter solutions first!

1. Rules, heuristics, good old code
2. Traditional machine learning - lots of training data
3. Small, specialized language models
4. General purpose LLMs

Rise of "small" language models (~10B or less)

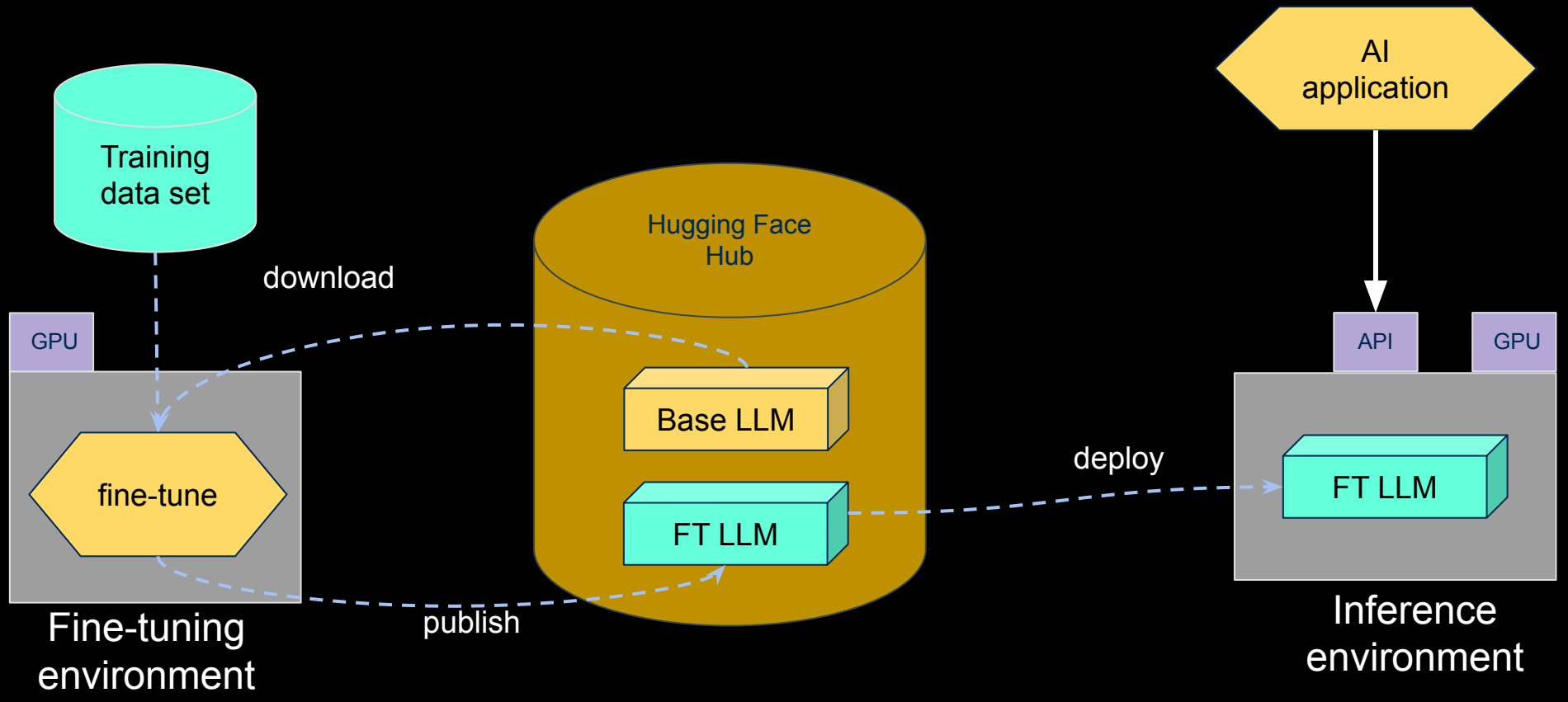
Smaller can be better and faster



Example from Predibase

3. Don't depend on corporate AI

DIY, modular AI systems development



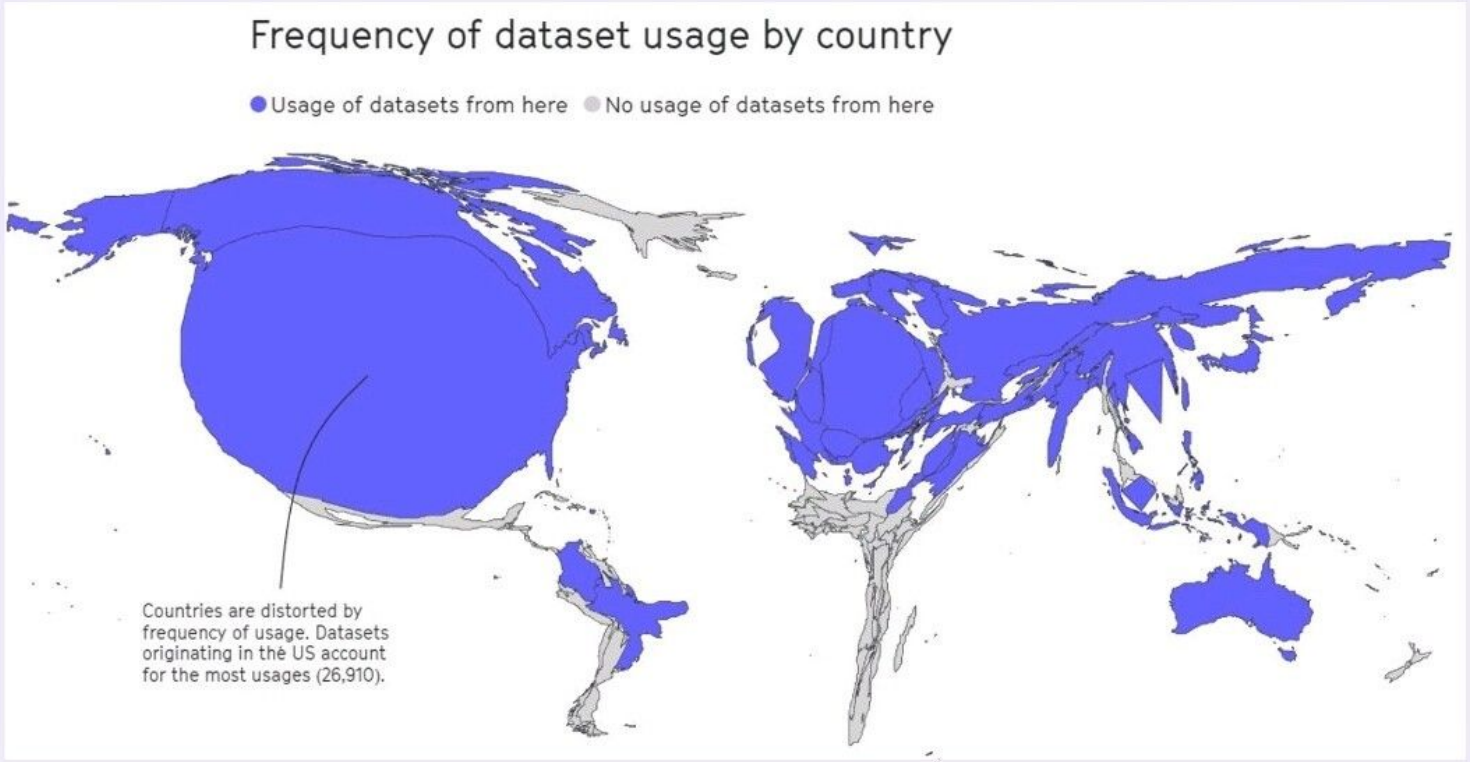
All pieces are interchangeable,
including the base model

4. Evaluate & Create data sets

Common questions about AI

- is it biased?
- does it hallucinate?
- can you trust it?
- does it know my language and culture?

The World Map according to the data AI sees



Sources

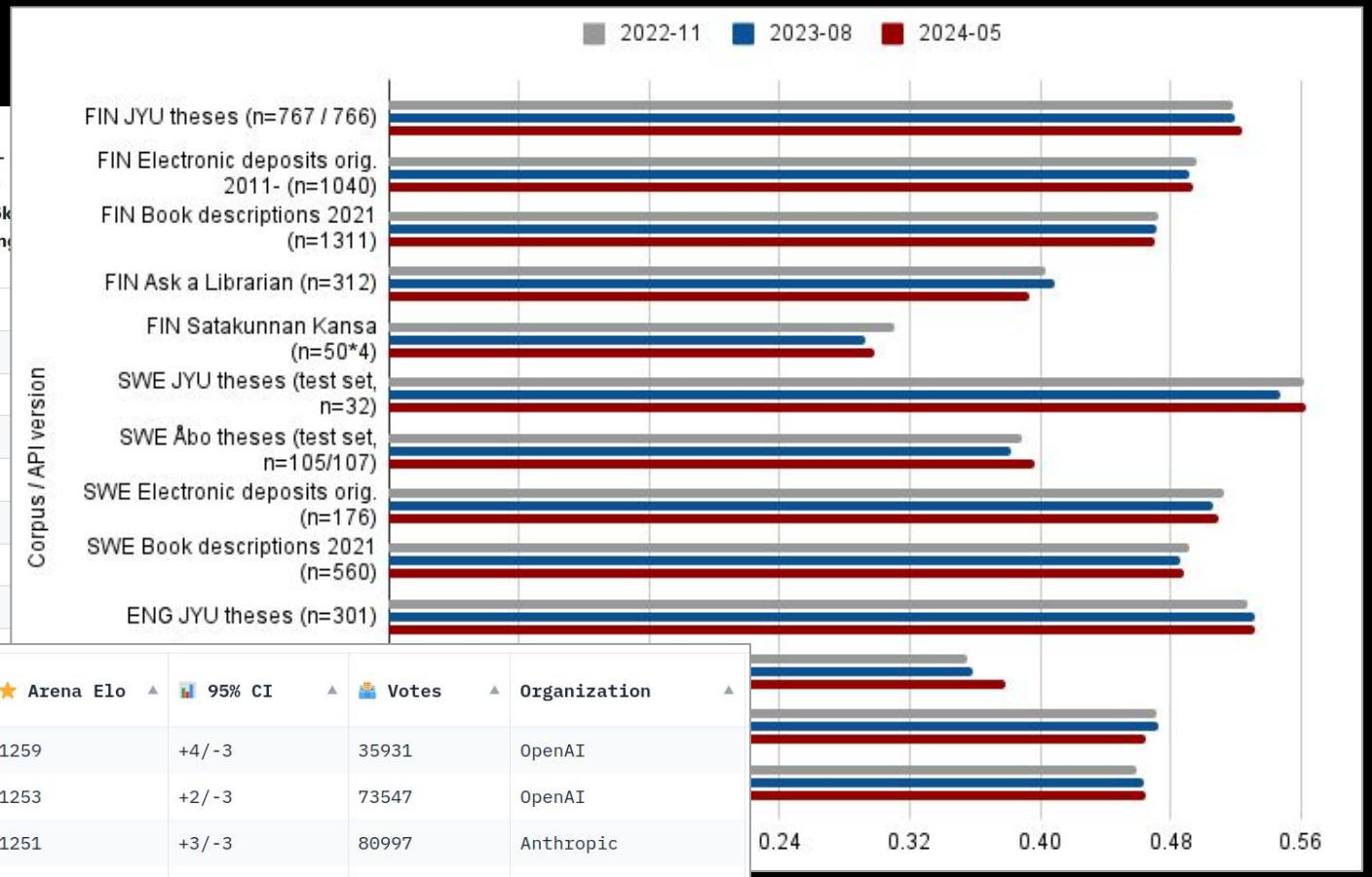
Research by: Koch, Denton, Hanna, and Foster (2021)

Visual by: The Mozilla Internet Health Report 2022

LLM Metadata Extraction model evaluation

language	field	axolotl-fine-tune-MiniChat-1	axolotl-fine-tune-Nous-Hermes-2-Mistral-7B-DPO	axolotl-fine-tune-stablelm-2-zephyr-1_6b	baseline-null	ludwig-fine-tune-zephyr-7b	meteor	openai-gpt35-turbo-16k prompting
eng	dc.contributor.author	0.7458	**0.8136**	0.7797	0.0508	0.7966	0.5763	0.7966
eng	dc.date.issued	0.9153	**0.9492**	0.8983	0	0.9322	0.7119	0.8136
eng	dc.identifier.isbn	0.6271	0.8305	0.8136	0.4746	**0.8814**	0.7966	0.6102
eng	dc.language.iso	0.8983	**1.0000**	0.9661	0	0.9831	**1.0000**	0.5254
eng	dc.publisher	0.5593	**0.7627**	0.6780	0.0169	0.6780	0.0508	0.5254
eng	dc.relation.eissn	0.7627	**0.9492**	0.9322	0.7288	0.9153	0.8475	0.8983
eng	dc.title	0.8644	**0.8983**	0.8475	0	0.8644	0.5763	0.8475
fin	dc.contributor.author	0.7538	**0.9077**	0.7692	0.2	0.7692	0.7077	0.7385
fin	dc.date.issued	0.9231	0.9231	0.923				
fin	dc.identifier.isbn	0.7077	**0.9077**	0.876				
fin	dc.language.iso	0.9846	0.9846	0.984				
fin	dc.publisher	0.6	**0.8615**	0.815				
fin	dc.relation.eissn	0.8	0.9385	0.907				
fin	dc.title	0.5538	0.7692	0.630				

Annif Subject Indexing evaluation

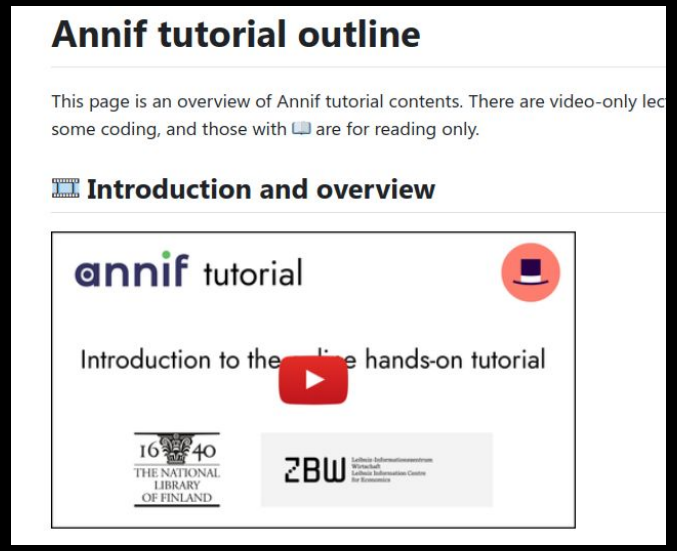


Rank* (UB)	Model	Arena Elo	95% CI	Votes	Organization
1	GPT-4-Turbo-2024-04-09	1259	+4/-3	35931	OpenAI
2	GPT-4-1106-preview	1253	+2/-3	73547	OpenAI
2	Claude 3 Opus	1251	+3/-3	80997	Anthropic
2	Gemini 1.5 Pro API-0409-Preview	1250	+3/-3	39482	Google
2	GPT-4-0125-preview	1247	+3/-2	67354	OpenAI
6	Llama-3-70b-Instruct	1210	+3/-4	53404	Meta
6	Bard (Gemini Pro)	1209	+5/-6	12387	Google
7	Claude 3 Sonnet	1201	+2/-3	78956	Anthropic
9	Command R+	1191	+3/-3	44988	Cohere
9	GPT-4-0314	1190	+3/-4	52079	OpenAI

LMSYS Chatbot Arena Leaderboard

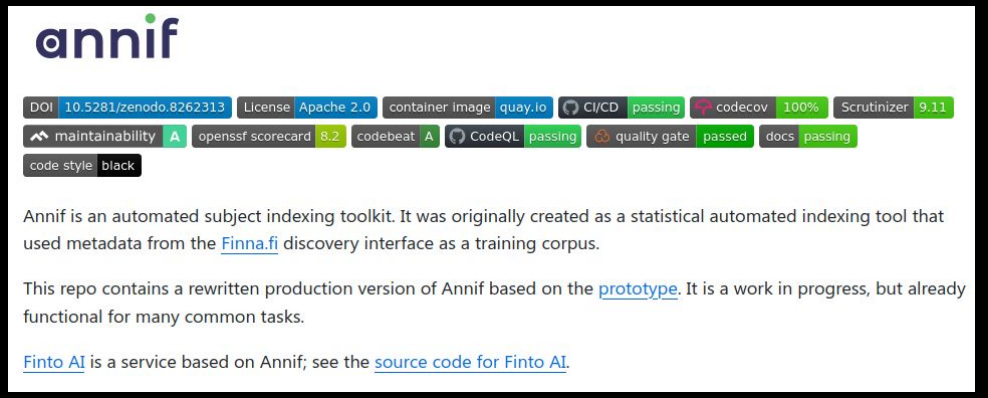
5. Be open and transparent

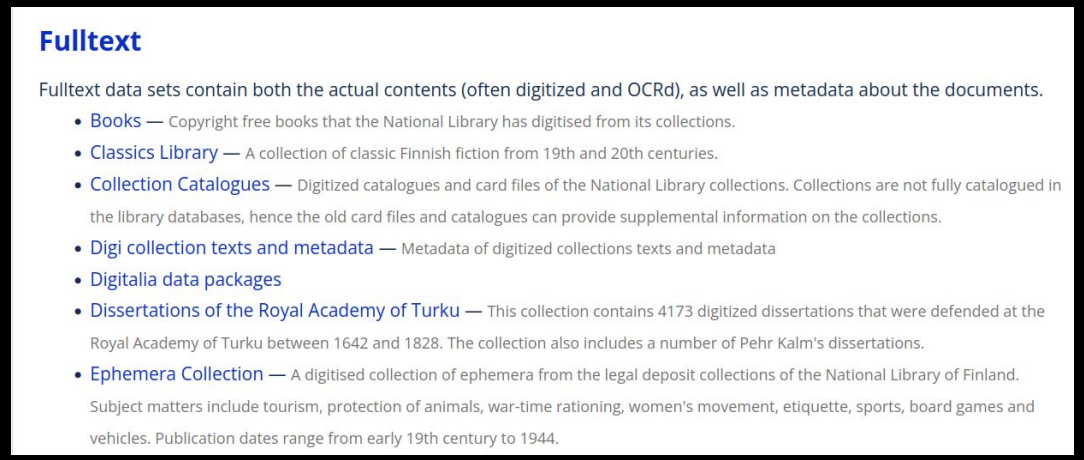
Open code, open models, open and inclusive communication



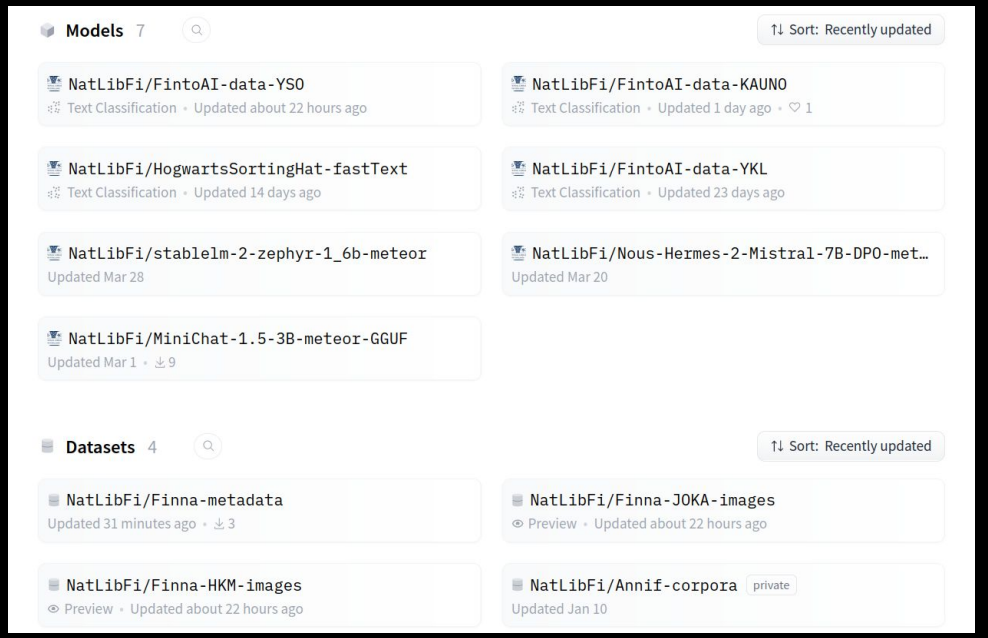
Annif tutorial

Annif code on GitHub





Datasets in NLF Data Catalog



Models and Datasets in Hugging Face Hub

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Thank you!

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Skosmos

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