

Building Open LLMs for Europe

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About Me



PhD



Research Engineer (now)



Researcher (until May 2024)



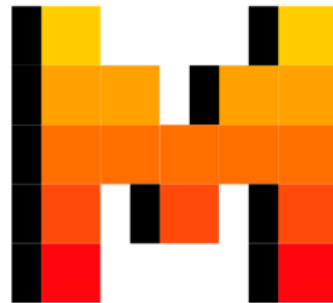
Open Legal Data



Occiglot



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Recap: Large Language Models

- Language models are **statistical models** that learn a probability distribution over a sequence of words from their training data and that predict **the next token with the highest probability** for a given input text.
- Tokenization converts natural language text into numerical vector representations based on a **fixed and limited vocabulary**.
- Transformer architecture allows the scaling of language models **in terms of parameters, data, and compute** (resource requirements).
- Large-scale of today's language **model enables generalization and solving of novel tasks** with no or little additional training data (few- and zero-shot).

How to build a large language model?

... in the open and for Europe.

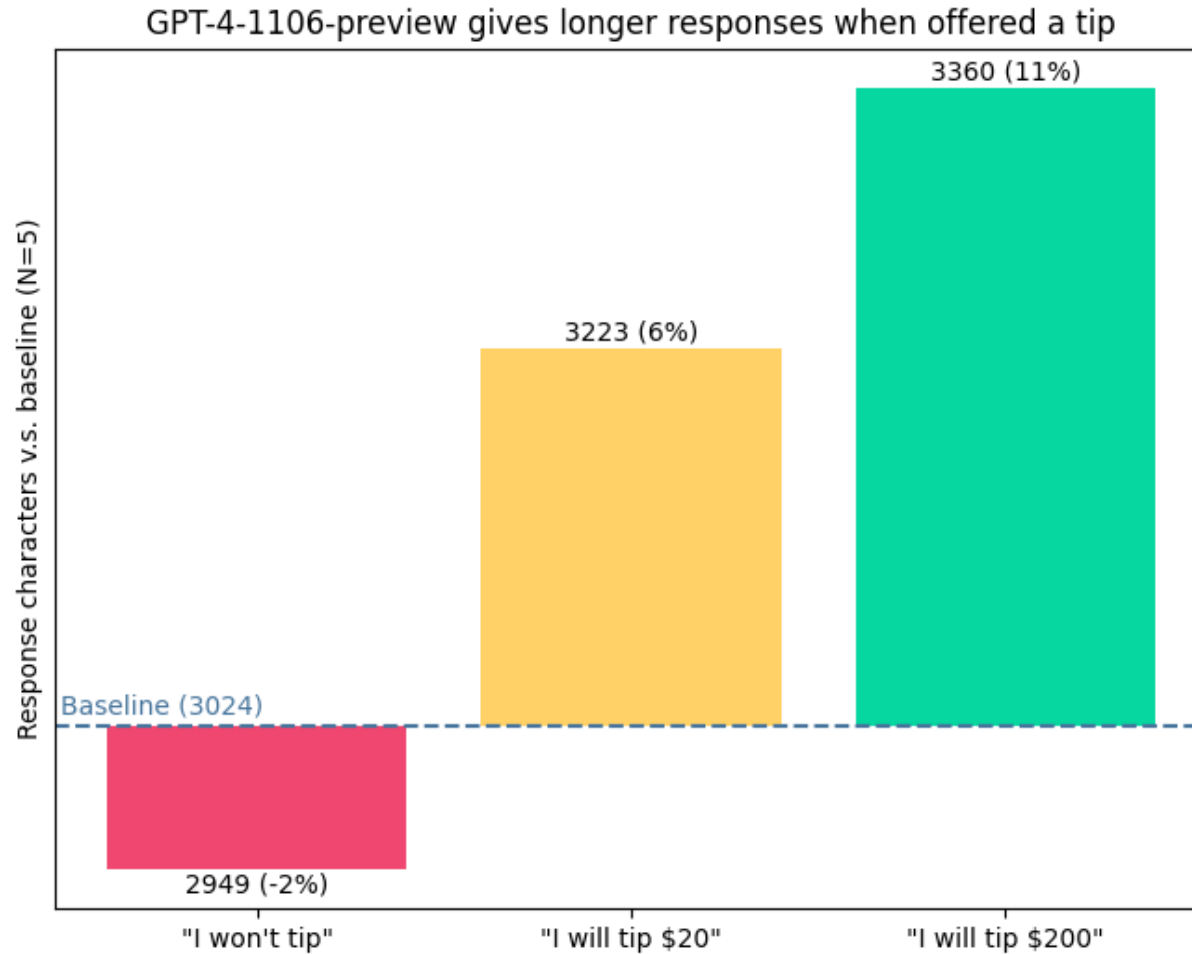
Open LLMs?

Open LLMs?

- “Open source doesn’t just mean access to the source code.” (Open Definition)
 - **Free Redistribution**
 - **Derived Works ...**
- Open Weights: The model weights are openly available, you can inspect them, and the model can be run on your own hardware - **but other license restrictions might apply (not truly open)**.
 - LLAMA2: only free use for services with < 700M monthly active users
 - Cohere Command R: non-commercial license (CC-BY-NC)
- Statistical models: “source code = training data“ ← **Our goal**

For Europe?

ChatGPT is American.



Source: <https://x.com/voooooogel/status/1730726744314069190> (@ voooooogel on X)

Tokenization

- Tokenization is the foundation of language models: Conversion of natural language text into tokens.
- Segmentation by different tokenizers: “zusammenarbeiten”
 - GPT4 tokenizer: [z] [us] [ammen] [arbeit] [nen]
 - German tokenizer: [zusammen] [arbeiten]
- Model costs (API-calls or compute time) are highly depended on the tokenization (number of tokens).
- Self-attention: quadratic complexity $O(n^2)$ with n tokens
- **Up to 68% more training costs** with suboptimal tokenizer.

Performance difference between the worst and best tokenizer:

	Task	Min	Max
English	ARC-Easy	0.50	0.59
	HellaSwag	0.34	0.41
	MRPC	0.54	0.69
Multilingual	XNLI FR	0.37	0.49
	XNLI EN	0.49	0.52
	X-CODAH ES	0.28	0.43
	10kGNAD	0.15	0.43

Publication: Ali et al., “Tokenizer Choice For LLM Training: Negligible or Crucial?” <https://arxiv.org/abs/2310.08754>

Data

Data matters!

"We find that data quality is critical to a highly-performing model"

(Google Gemini technical report, 2023)

"Data curation was the most important work for building Grok"

(Elon Musk at the Lex Friedman Podcast, 2023)

Training data

Stage 1: Unsupervised Pretraining

- Large amounts of plain text (Llama3: **15 trillion tokens ~ 500k years of human typing**)
- Diverse sources and topics (scientific literature, news, forums)
- Most prominent source: Web crawled text (CommonCrawl)

Stage 2: Supervised Fine-tuning

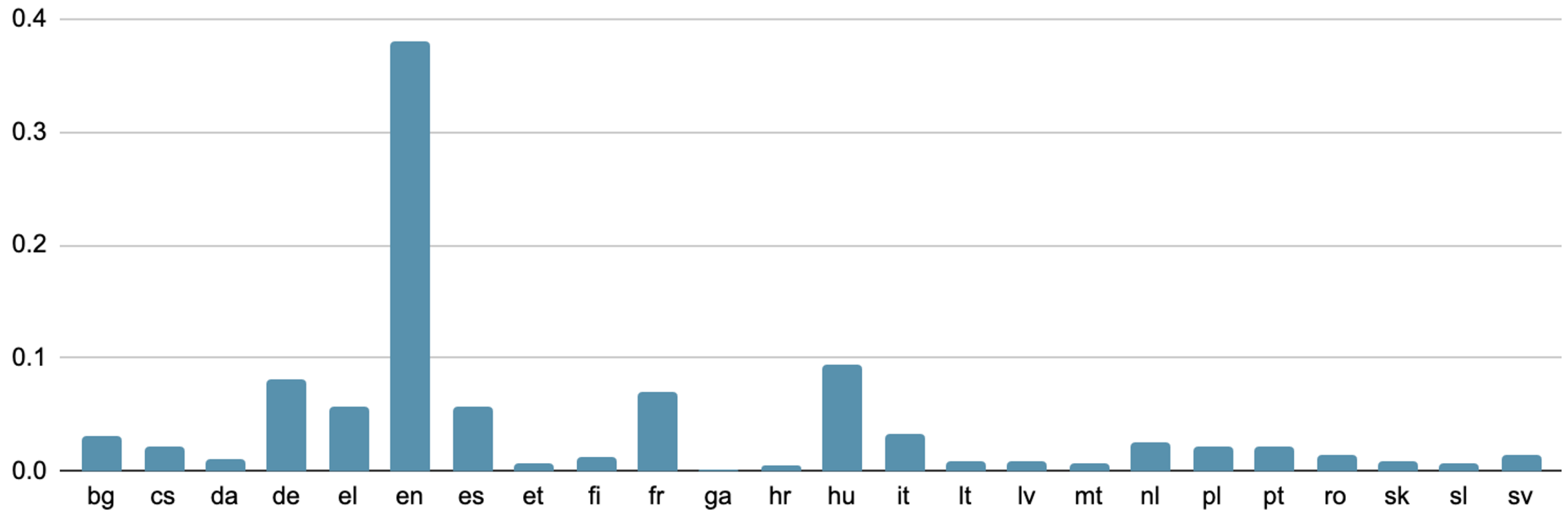
- Supervised text pairs (input-output, question-answer, text-summary)
- Task-oriented data (diverse tasks are needed to generalize to unseen tasks)

Stage 3: Alignment & preference-tuning

- Human (or AI) feedback data on preferred output
- Pairwise feedback (good vs bad)
- Listwise feedback (ranking)



European LLM Data?



Available pretraining data by language based on OSCAR v23.01



Where is the data coming from?

- The only source that provides **enough data at low costs** is the Web.
- [CommonCrawl](#): US-based non-profit that crawls the Web
 - 250 billion Web pages spanning 17 years (petabytes of data)
 - CC-Crawler operates with an US-IP address and an English user agent.
 - [OpenWebSearch](#): Initiative for building a European Web search infrastructure.
- In addition to Web-crawled data, smaller but higher quality datasets are used (curated dataset such as scientific literature, news, ...).



LLM-Datasets

- **LLM-Datasets** is a collection of datasets for language model pretraining including scripts for downloading, pre-processing, and sampling.
- Datasets for +32 European languages available
- Filtered text data: approx. 2 Trillion Tokens (comparable to LLama2)
- Easy to extend with your own datasets without the need of making your data publicly available.



github.com/malteos/llm-datasets

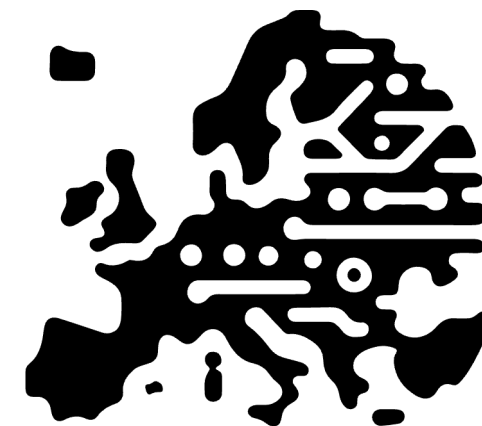
Apache 2.0 license

Preprint: Malte Ostendorff, Pedro Ortiz Suarez, Lucas Fonseca Lage, and Georg Rehm. LLM-Datasets: An Open Framework for Pretraining Datasets of Large Language Models. <https://ostendorff.org/assets/pdf/ostendorff2024-preprint.pdf>

Community-driven LLM development

Occiglot: Open Language Models for Europe

- Most LLMs are primarily trained and optimized for English, leading to lower performance and higher costs for other languages.
- To change this, we started **Occiglot** - A large-scale research collective for open-source development of Large Language Models by and for Europe.
- Community-driven effort to make the LLM technology available for European languages (no official research project).
- **Model release v0.1:**
 - Continued pretraining and instruction-tuning based on Mistral 7B
 - Top-5 EU-languages: English, French, German, Spanish, and Italian
 - Bilingual (English + X) and multilingual models (Apache 2.0 license)
 - More languages are work-in-progress: Dutch, Portuguese, ...
- Released last Saturday: **Llama3-8B-DiscoLM-German**



Evaluation

Evaluation: Italian benchmarks

Model	Avg.	ARC IT	TruthfulQA IT	Belebele IT	HellaSwag IT	MMLU IT
Mixtral-8x22B-v0.1	66.9	66.1	28.7	88.8	79.5	71.4
Llama-3-SauerkrautLM-8b-Instruct	60.8	61.9	31.0	83.3	70.3	57.5
Spaetzle-v60-7b	59.9	59.3	34.6	81.7	69.1	54.8
llama3-8b-spaetzle-v20	59.8	59.7	29.6	83.9	67.9	58.0
occiglot/occiglot-7b-it-en-instruct	56.1	54.6	30.4	71.8	71.4	52.3
Meta-Llama-3-8B	55.6	50.3	26.4	80.0	65.4	55.9
Llama3-DiscoLeo-Instruct-8B-v0.1	54.5	49.3	31.3	77.4	63.2	51.4
Llama3-DiscoLeo-Instruct-8B-32k-v0.1	54.3	48.9	32.1	76.2	63.1	51.4
Mistral-7B-Instruct-v0.2	54.2	51.9	35.0	70.3	63.9	49.9

Occiglot Euro LLM Leaderboard



The Occiglot euro LLM leaderboard evaluates a subset of the tasks from the [Open LLM Leaderboard](#) machine-translated into the four main languages from the [Okapi benchmark](#) and [Belebele](#) (French, Italian, German and Spanish).

The translated tasks are uploaded to a fork of the great [Eleuther AI Language Model Evaluation Harness](#).

Disclaimer: A language is not represented by a country. Different languages can be spoken and spread in all countries around the globe. For the sake of simplicity, we have used flag emojis (🇫🇷 🇮🇹 🇪🇸 🇬🇧) to represent the language, not the countries.

LLM Benchmark About Submit here! Support us

Search for your model (separate multiple queries with `;`) and press ENTER...

Select columns to show

Average 🇩🇪 DE 🇫🇷 FR 🇮🇹 IT 🇪🇸 ES 🇬🇧 EN 🇬🇧 ARC EN 🇬🇧 TruthfulQA EN 🇬🇧 Belebele EN 🇬🇧 HellaSwag EN

🇩🇪 MMLU EN 🇩🇪 ARC DE 🇩🇪 TruthfulQA DE 🇩🇪 Belebele DE 🇩🇪 HellaSwag DE 🇩🇪 MMLU DE 🇫🇷 ARC FR 🇫🇷 TruthfulQA FR

🇫🇷 Belebele FR 🇫🇷 HellaSwag FR 🇫🇷 MMLU FR 🇫🇷 ARC IT 🇫🇷 TruthfulQA IT 🇫🇷 Belebele IT 🇫🇷 HellaSwag IT 🇫🇷 MMLU IT 🇪🇸 ARC ES

🇪🇸 TruthfulQA ES 🇪🇸 Belebele ES 🇪🇸 HellaSwag ES 🇪🇸 MMLU ES

Show gated/private/deleted models

Model	Average	🇩🇪 DE	🇫🇷 FR	🇮🇹 IT	🇪🇸 ES	🇬🇧 EN
mistral-community/Mixtral-8x22B-v0.1	68.3	66.81	67.48	66.9	61.52	72.87
VAGOsolutions/Llama-3-SauerkrautLM-8b-Instruct	64.49	60.07	63.05	60.81	57.91	74.71
mistralai/Mistral-7B-Instruct-v0.2	57.96	53.52	57.58	54.22	52.82	67.63
meta-llama/Meta-Llama-3-8B	57.25	53.89	56.68	55.59	50.92	63.08
mistralai/Mistral-7B-v0.1	56.03	52.8	55.81	53.34	50.66	62.73

<https://hf.co/spaces/occiglot/euro-llm-leaderboard>

Multilingual benchmarks: Lost in translation

	Occiglot-7B-EU5				Mistral-7B-v0.1		
Translation/prompt	ARC-DE	Hellaswag-DE	MMLU-DE		ARC-DE	Hellaswag-DE	MMLU-DE
Okapi (EN prompts)	0.494	0.667	0.483		0.476	0.610	0.527
Okapi (DE prompts)	0.489	0.667	0.487		0.483	0.489	0.524
LeoLM	0.491	0.647	0.485		0.524	0.588	0.473



Using different translations and prompts leads to different scores!

Evaluation: Human verification

Model	Translation quality
wmt21	0.848
GPT4	0.846
Claude-3-Opus	0.846
deepl	0.844
GPT3.5	0.844
Occiglot-DE-EN-Instruct	0.831
discolm	0.831
nbbl	0.829
wmt19	0.825

Community contribution!

https://github.com/CrispStrobe/llm_translation

Join the Occiglot community



Open Weekly Meeting
Every Tuesday 10am CEST



<https://occiglot.eu>

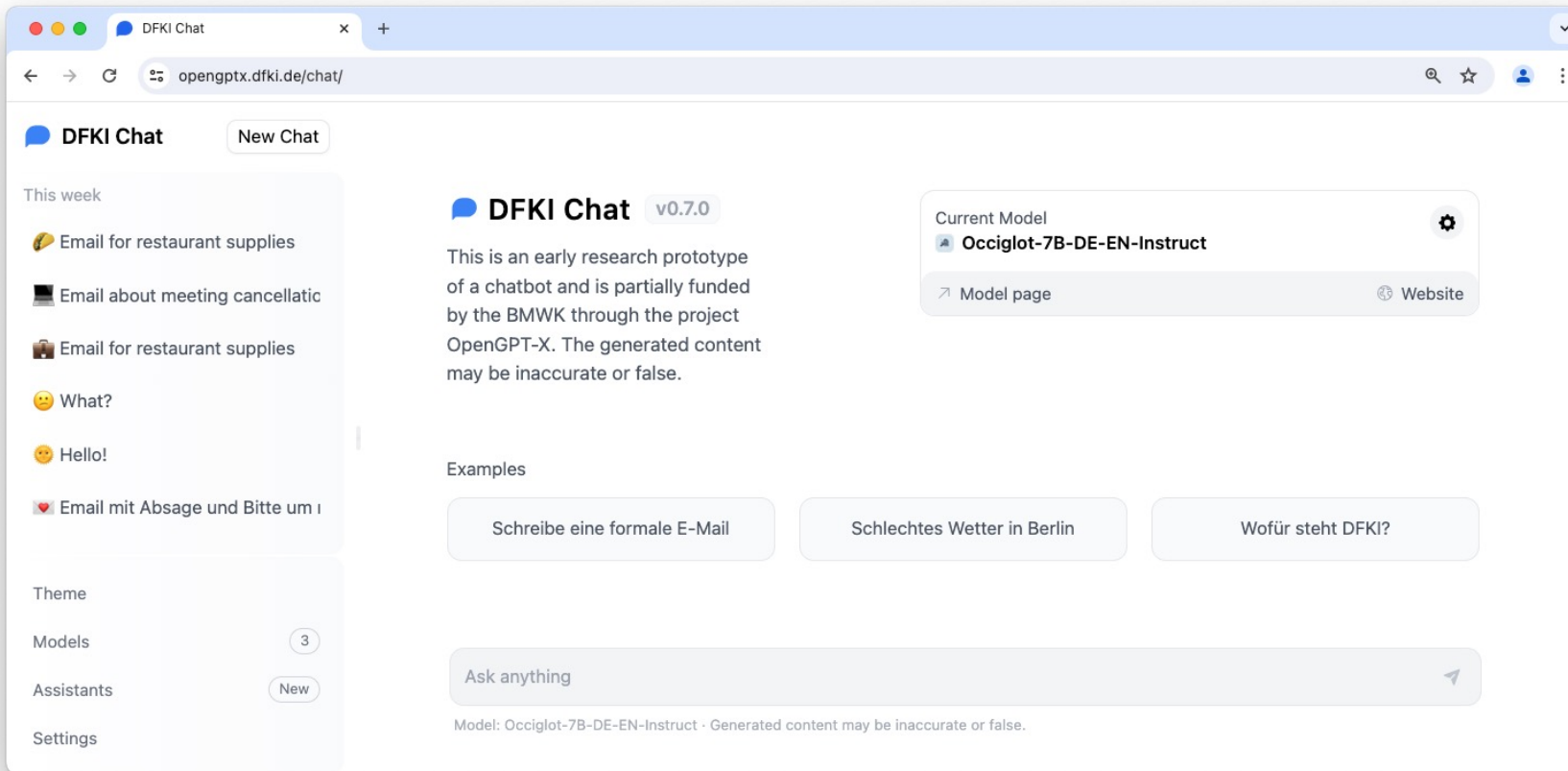
Web Data Curation

- Web data is noisy and often of bad quality and thus harming model performance.
- Improvements of Web data quality will have a large and long-lasting impact on model performance.
- We are collecting information about “good” and “bad” domains for better filtering of Web data.
- Collaboration with CommonCrawl: more crawling of good domains (used by all major LLM providers)
- Required skills: “**Web understanding**”
- Task: Add domains to our spreadsheet

<https://github.com/occiplot/curated-web-data>

Top Web domains from Clean Colossal OSCAR 2323-IT:

domain	chars
englishgratis.com	213950085
www.oranews.net	152192225
stefanocipolla.com	120118131
www.camera.it	74776273
curia.europa.eu	66268691
www.sdb.org	65471927
progettogayforum.altervista.org	59241844
ilmanifesto.it	58036699
www.medicalsportsrl.it	46686423
demetra.regione.emilia-romagna.it	46066548
leg16.camera.it	42917806
www.ilmio psicologo.it	42657780
www.storiologia.it	41275056



<https://opengptx.dfki.de/chat/>

Thank you! Any questions?



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[@xyou](#)



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