MaCoCu: Massive collection and curation of monolingual and bilingual data: focus on under-resourced languages

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Abstract

We present the most relevant results of the project *MaCoCu:* Massive collection and curation of monolingual and bilingual data: focus on under-resourced languages in its second year. Parallel and monolingual corpora have been produced for eleven low-resourced European languages by crawling large amounts of textual data from selected top-level domains of the Internet; both human and automatic evaluation show its usefulness. In addition, several large language models pretrained on MaCoCu data have been published, as well as the code used to collect and curate the data.

1 Introduction

This paper describes the main outcomes of the project *MaCoCu: Massive collection and curation of monolingual and bilingual data: focus on underresourced languages* (Bañón et al., 2022), spanning from June 2021 to July 2023. MaCoCu is aimed at building large and high-quality monolingual and parallel (with English) corpora for ten low-resourced European languages (see Table 1). The international consortium behind this project consists of four partners: Jožef Stefan Institute (Slovenia), Rijksuniversiteit Groningen (Netherlands), Prompsit Language Engineering S.L. (Spain), and Universitat d'Alacant (Spain; coordinator).

Other existing initiatives, such as Paracrawl¹ or Oscar² exploit existing resources such as Common Crawl³ or the Internet Archive.⁴ Our strategy con-

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sists in automatically crawling top-level domains (TLD), potentially containing substantial amounts of text in the targeted languages,⁵ and then applying a monolingual and a parallel curation pipelines. The evaluation of the first data release (van Noord et al., 2022a) confirms the usefulness of these data for different natural-language processing tasks.

2 Collected corpora

Monolingual and parallel corpora are built from crawled data by applying a thorough cleaning process, including noise fixing/filtering and removal of near-duplicate/boilerplate text. Corpora are then automatically annotated with: (a) document and paragraph IDs; (b) language variety (e.g. British/American English); (c) document-level affinity to DSIs identified through domain modelling (van Noord et al., 2022b); (d) personal information; and (e) identification of translated text: either human or machine translations (only for parallel corpora). Table 1 shows the size of the corpora for the second data release, published in April 2023.

2.1 Data evaluation

To the date, evaluation only covers the seven languages included in the first data release of the action, made public in April of 2022.

Mono-lingual A set of pre-trained language models (LMs)⁶ has been built and released for Icelandic, Maltese and Bulgarian/Macedonian by continuing the training of multilingual XLM-RoBERTa-large (Conneau et al., 2020) using only MaCoCu data for all languages. These models outperform monolingual baselines, and XLM-R and large models on the POS, NER and COPA (Roemmele et al., 2011)

https://paracrawl.eu/

²https://oscar-project.org/

³https://commoncrawl.org/

⁴https://archive.org/

⁵National TLDs such as .hr for Croatian, or .is for Icelandic, and also generic TLDs such as .com, .org, or .eu. ⁶https://huggingface.co/MaCoCu

	Monolingual		Parallel	
Language	Docs.	Words	Segs.	Words
Turkish	16.0	4344.9	1.6	89.2
Bulgarian	10.5	3506.2	1.8	72.1
Croatian	8.1	2363.7	2.3	99.5
Slovenian	6.3	1920.1	1.9	85.0
Macedonian	2.0	524.1	0.4	18.3
Icelandic	1.7	644.5	0.3	10.6
Maltese	0.5	347.9	0.9	53.9
Albanian	1.7	625.7	0.5	24.3
Serbian	7.5	2491.0	2.1	95.9
Montenegrin	0.6	161.4	0.2	11.2
Bosnian	2.8	730.3	0.5	22.2

Table 1: Sizes for corpora in the 2nd data release. Monolingual corpora are measured in millions of documents (Docs.) and millions of words. Parallel corpora are measured in millions of parallel segments (Segs.) and millions of words. Bosnian is a bonus language as it was not initially covered in the action.

	bg	is	mk	mt	tr
XLM-R-base	56.9	55.2	55.3	52.2	53.2
XLM-R-large	53.1	54.3	52.5	54.0	50.5
Monolingual LM	_	54.6	_	55.6	56.4
XLM-R + MaCoCu	54.6	59.6	55.6	54.4	58.5

Table 2: Test set COPA scores for baseline LMs compared to continuing training XLM-R-large on MaCoCu data.

evaluation tasks. Table 2 shows the results for the COPA test set, the most challenging evaluation task. For Bulgarian/Macedonian we also train an LM from scratch using the RoBERTa (Liu et al., 2019) architecture, dubbed BERTovski, which reached competitive performance with XLM-R.

Parallel Parallel data were extrinsically evaluated first training neural machine translation systems on large data sets available on OPUS⁷ (ParaCrawl, CommonCrawl, Tilde), and comparing the results obtained when adding the MaCoCu data to the training set. Results show improved performance for all languages across different evaluation sets and metrics. These results were confirmed by human evaluation (van Noord et al., 2022a).

3 Free/open-source pipeline

The curation pipelines used to produce MaCoCu corpora, Monotextor⁸ and Bitextor,⁹ have been re-

leased under free/open-source licences. Crawling and corpora-enrichment software have been also released under the MaCoCu¹⁰ GitHub organisation.

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⁷https://opus.nlpl.eu/

⁸https://github.com/bitextor/monotextor

⁹https://github.com/bitextor/bitextor

¹⁰https://github.com/macocu