

Learning From Arabic Corpora But Not Always From Arabic Speakers: A Case Study of the Arabic Wikipedia Editions

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Outline

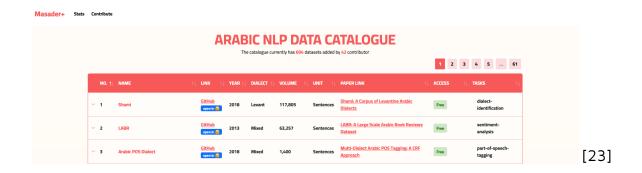
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Abstract

- We present a case study focusing on differences among the Arabic Wikipedia editions (Modern Standard Arabic (MSA), Egyptian, and Moroccan).
- We document issues in the Egyptian Arabic Wikipedia with automatic creation/generation and translation of articles from English without human intervention.



- NLP is a key element in decision-making systems like resume parsers that sort lists of job candidates.
- These systems **need** large corpora of human text to understand humans to make recommendations [1].



Background

- These corpora of human text **convey** many social concepts, including culture, traditions, perspectives, and even historic biases [2, 3, 4, 5].
- Yet, having a corpus of text in a language does not necessarily represent the culture of native speakers of that language.

Background

- Human languages are under-represented in **both** corpora development and NLP toolchain support [1].
- This under-represents the culture and views of native speakers of those languages in NLP-guided decisionmaking.

Motivation

- Human corpora **should** be written by native speakers, yet others may be written by non-native speakers or translated from other languages [6].
- Such unrepresentative corpora could affect the performance of many NLP tasks such as Large Language Models (LLMs) trained from these corpora.

Motivation

- Wikipedia corpora (content pages) are widely **used** in NLP to train LLMs like ELMo, BERT, and GPT-3 [7, 8, 9].
- Some Wikipedia corpora have been developed/created through bots or automated scripts; this often involves translation from other languages [10].

Motivation

 This work highlights the differences between text corpora written by native speakers and those translated and generated by automated systems.

- We compare the Arabic Wikipedia editions regarding pages to date, new pages, and top editors, besides English Wikipedia as a benchmark.
 - We took a data snapshot of the **four** Wikipedia editions' statistics in July 2022 using the online Wikimedia Statistics service (https://stats.wikimedia.org).



 We contribute our implementation of the Wikimedia Statistics service as a Python package and Command Line Interface. Wikistats-to-CSV (wikistats2csv): https://github.com/SaiedAlshahrani/Wikistats-to-CSV.

```
>>> from wikistats2csv import Content
>>> Content.pages_to_date(wiki='es', period='all-years', filter='page-type-all', interval='monthly')
## Downloaded `spanish--pages-to-date--page-type-all--all-years--monthly.csv` successfully :-)
** Quick glance at `spanish--pages-to-date--page-type-all--all-years--monthly.csv` file:
                       month total.non-content total.content
                                                                         timeRange.start
                                                                                                     timeRange.end
    2001-01-01T00:00:00.000Z
                                                                2001-01-01T00:00:00.000Z 2001-02-01T00:00:00.000Z
    2001-02-01T00:00:00.000Z
                                                                2001-02-01T00:00:00.000Z 2001-03-01T00:00:00.000Z
257 2022-06-01T00:00:00.000Z
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                                        3896209
258 2022-07-01T00:00:00.000Z
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```
$ wikistats2csv -w ar -m content -q pages-to-date -p all-years -f page-type-all -i monthly
            ## Downloaded `arabic--pages-to-date--page-type-all--all-years--monthly.csv` successfully :-)
** Quick glance at `arabic--pages-to-date--page-type-all--all-years--monthly.csv` file:
                     month total.non-content total.content
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    2001-01-01T00:00:00.000Z
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```

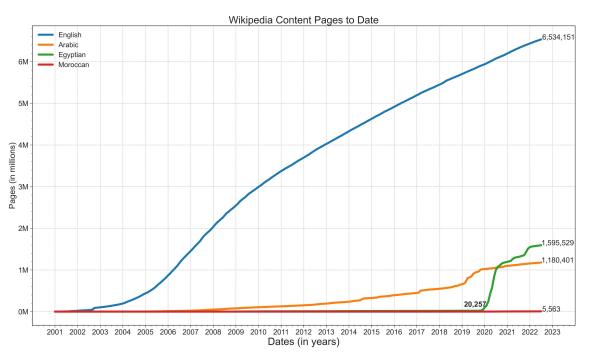
- We study the impact of problems in Egyptian Arabic Wikipedia, including large-scale automatic generation and poor translation of content pages from English.
 - Interestingly, over 63% of the total content pages in Egyptian Wikipedia have been automatically created using template-based translation from English.

The Case of Wikipedia Arabic Wikipedia Editions

- Wikipedia encyclopedia was launched 20 years ago (in 2001) and was released primarily in English.
- The Arabic, Egyptian, and Moroccan editions appeared on the project in 2004, 2008, and 2019, respectively.

Language	Code	Articles	Total Pages	Edits	Admins	Registered Users	Active Users
English	en	6,543,738	56,401,668	1,101,698,546	1,032	44,056,435	114,504
Arabic	ar	1,183,778	7,815,021	58,966,845	26	2,293,115	4,820
Egyptian Arabic	arz	1,596,851	2,010,972	7,343,259	7	189,191	190
Moroccan Arabic	ary	5,744	43,714	188,790	3	6,415	31

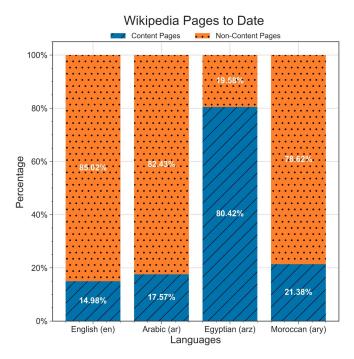
- English, Arabic, and Moroccan Arabic show normal growth in their content pages (articles) over the timeline of Wikipedia.
- The content of Egyptian Arabic has recently grown rapidly and exponentially in the last two years.



Wikipedia	Articles	Period	Monthly	Daily
Arabic	~1.2 million	19 years	~5,000	~200
Egyptian	~1.6 million	Less than 3 years	~50,000	~2,000

 This exponential growth of the content pages in the Egyptian Arabic Wikipedia in only 30 months is the result of the large-scale automated creation of the content pages.

 We visualize the percentage of all page types to date for the **four** Wikipedia editions, displaying the difference in percentage between page types to study the characteristics of each Wikipedia within itself.

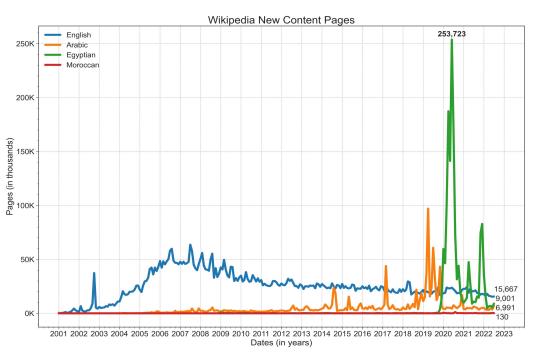


 Egyptian Arabic Wikipedia has approximately 20% of non-content pages and 80% of content pages, which is a consequence of the large-scale automation of content creation.

The Case of Wikipedia Arabic Wikipedia Editions: New Pages

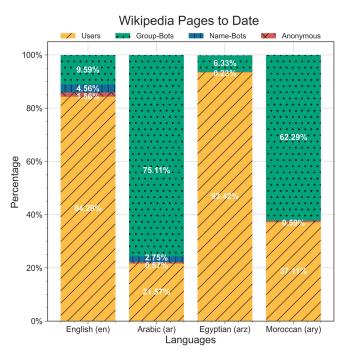
- In June 2020, approximately **253,000** new content pages were created in the Egyptian Arabic Wikipedia.
- Nearly 23,700 new content pages were created on English Wikipedia, nearly 4,280 on Arabic Wikipedia, and nearly 50 on Moroccan Wikipedia, all in June 2020.

The Case of Wikipedia Arabic Wikipedia Editions: New Pages



- Wikipedia has four types of editors:
 - Registered Users:
 - logged-in users but not in group-bot nor name-bot sets.
 - O Group-bots:
 - logged-in users who are part of a bot group.
 - O Name-bots:
 - logged-in users whose name contains 'bot'.
 - Anonymous Users:
 - unlogged-in users (tracked by IPs or devices fingerprints).

 We visualize the percentage of all pages to date for the four Wikipedia editions by displaying the difference in percentage between **editor types** to study the characteristics of each Wikipedia within itself.

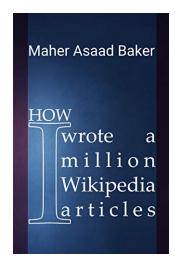


 We hypothesize that the reason behind the 84% of pages created by registered users in English Wikipedia is the huge gap in the total number of registered users (~233X bigger than Egyptian Wikipedia).

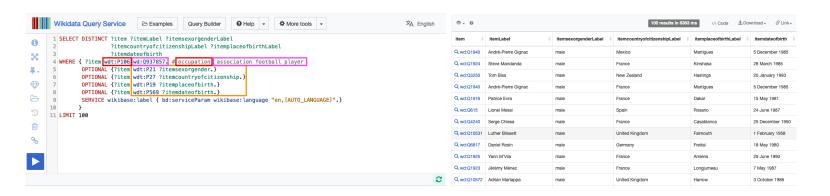
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- We found that over one million articles (63% of the total articles) in Egyptian Arabic Wikipedia have been created by one registered user called "HitomiAkane".
- This user has made more than 1,562,615 new creations, made nearly 1,615,216 edits, and created over 1 million auto-generated articles without human revision [11].

 This user described the large-scale content creation process using template-based translation from English in his book: How I Wrote a Million Wikipedia Articles [10].









 We quote the example of football player that he used to demonstrate the automation process in the book:

[label] [date of birth], [gender] is a football player from [citizenship], [gender] was born at [date of birth] in [place of birth].

- The process of converting the translated MSA articles to the Egyptian Arabic dialect remains **mysterious**.
 - We hypothesize that the user used a lexicon-based conversion between MSA and Egyptian to make it look like it was produced organically by native speakers.
- Overall, the used process represents a relatively shallow, template-based content translation.

- Wikipedia articles should only be written, contributed to, edited, and maintained by the people.
- This lack of representativeness and cultural richness holds many potential problems that could negatively impact society when deploying AI systems or NLP tools trained on such unrepresentative corpora [12].

Persistent Anti-Muslim Bias in Large Language Models [22]

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Discussion

- Arabic poses many **challenges** in NLP that prevent simply translating from another language due to its morphological richness and high ambiguity [13, 14].
 - Arabic verbs have ~5,400 inflected forms, English verbs have 6, and Chinese verbs have only one [15].
 - Diacritical marks vs. multiple variants of a word [5].

Discussion

 Arabic has many dialectal variants, like Egyptian and Moroccan Arabic, that are different from MSA, which are primarily spoken, do not have orthographic rules, and have few resources [16, 17].

Discussion

 Despite all these challenges, translating from English to enrich low-resource languages' content like Arabic is a **common** practice, mainly done using Machine Translation models (MTs) [18].

Representativeness

- The heart of the lack of representativeness problem, specifically in the Arabic language, can be discussed from two different perspectives:
 - 1. The template-based large-scale auto-generation of content, especially in the Wikipedia encyclopedia.
 - 2. The translation of content from English to Arabic using direct translation methods or off-the-shelf tools.

Representativeness: Generation

 We have documented that the Egyptian Arabic Wikipedia content **does not** genuinely represent the Egyptian people, their culture, traditions, or views.

Representativeness: Generation

 We, humans, express our values, traditions, or perspectives through writing, and when using autogenerated text corpora, we **miss** the people's culture, sentiments, stances, opinions, etc.

Representativeness: Generation

- Even though Wikipedia articles are factual entries, the choice of topics, choice of facts, and choice of words reflect the perspective and values of the authors.
- It **matters** whether these choices are made by native speakers or by translation from other languages.

Representativeness: Translation

 The other face of the lack of representativeness problem is **translating** the content of the English language to other low-resource languages like Arabic using direct translation or off-the-shelf tools.

Representativeness: Translation

 Wikimedia Foundation has encouraged users and contributors to use MTs to translate and create the initial content of more than 400,000 articles on Wikipedia project using Google Translate [19].

Representativeness: Translation

 Yet, we **should** consider the quality of these translation tools, the quality of the translation work conducted by non-expert Wikipedia users, and what they could bring to the content of Wikipedia from potential issues like biases and sexism [20, 21].

ACADEMIA Letters

Gender bias in machine translation: an analysis of Google Translate in English and Spanish

Discussion Final Statement

 NLP corpora that are automatically created using shallow templated-based, poorly translated using direct translation, auto-generated by advanced LLMs, or even the assembled corpora using text augmentation techniques do not echo the complex structure of the Arabic language and its dialects, do not express the views of the Arabic speakers, and do not represent the cultural richness and historical heritage of the Arabic language and its people.



- We studied the three Arabic Wikipedia editions besides English Wikipedia and shed light on the problem of the Egyptian Arabic Wikipedia.
 - We found that **one** registered user has automated the creation of over 1 million articles in less than three years and used a shallow, template-based translation method.

Recommendation

- We recommend that NLP practitioners avoid the inorganic, unauthentic, unrepresentative corpora in their applications when the goal is to learn from past human behavior and to investigate how the corpora they do use were created, generated, or assembled.
- We recommend that when registered users employ automated translation processes, their contributions should be marked differently than "registered user"; perhaps "registered user (automation-assisted)".

Future Works

- We plan to construct a representative analysis of the Arabic Wikipedia editions in terms of corpus and similarity metrics (work in progress).
- We plan to study the **implications** of using such unrepresentative corpora that are naively auto-created, poorly translated, or automatically generated on the downstream applications of the NLP.

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