



Limitations of Religious Data and the Importance of the Target Domain: Towards Machine Translation for Guinea-Bissau Creole

Jacqueline Rowe, Edward Gow-Smith, Mark Hepple







>1.5 million
L2 speakers

>350k
L1 speakers



Christoph Kohl. 2016. Limitations and Ambiguities of Colonialism in Guinea-Bissau: Examining the Creole and "Civilized" Space in Colonial Society. *History in Africa*, 43:169–203.

Kiriol	English
Timótiu mistiba juda jintis	Timothy wanted to help people
Sangi ta kuri na arteria.	Blood flows in the artery.
Un ermon sta ku si fiju, i na piskaba.	A brother was fishing with his son.

Kiriol	Portuguese
Timótiu mistiba juda jintis	Timóteo queria ajudar os outros.
Sangi ta kuri na arteria.	O sangue corre na artéria.
Un ermon sta ku si fiju, i na piskaba.	Um irmão estava pescando junto com seu filho.

Kiriol	Portuguese
Timótiu mistiba juda jintis	Timóteo queria <mark>ajudar</mark> os outros.
Sangi ta kuri na arteria.	O sangue corre na artéria.
Un ermon sta ku si fiju, i na piskaba.	Um irmão estava pescando junto com seu filho.

Kiriol	Portuguese
Timótiu misti ba juda jintis	Timóteo quer ia ajud ar os outros.
Sangi ta kuri na arteria.	O sangue corre na artéria.
Un ermon sta ku si fiju, i na piskaba.	Um irmão estava pescando junto com seu filho.

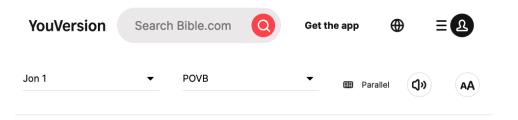
Creole NLP

- Work on individual creole languages
- Robinson et al. (2024). Kreyòl-MT: Building MT for Latin American,
 Caribbean and Colonial African Creole Languages.
- Lent et al. (2024). CreoleVal: Multilingual multitask benchmarks for creoles.

Creole NLP

- Work on individual creole languages
- Robinson et al. (2024). Kreyòl-MT: Building MT for Latin American,
 Caribbean and Colonial African Creole Languages.
- Lent et al. (2024). CreoleVal: Multilingual multitask benchmarks for creoles.
- MT performance shown to depend on vocabulary overlap (Birch et al 2008) and morphological complexity (*Koehn, 2005; Park et al., 2021; Cotterell et al., 2018; Arnett & Burgen (2024)*

Machine translation for Guinea-Bissau Creole (Kiriol)



JON 1

Palabra di Deus

Antis di mundu kumpudu, kil ki Palabra i tenba

ja; i staba ku Deus, i seduba mesmu Deus. ² El i staba
ku Deus na kumsada. ³ Deus fasi tudu kusa pa meiu
del; i ka ten nada ku fasidu sin el. ⁴ I nel ku vida staba;
ki vida i seduba lus di omis. ⁵ Lus numia na







Janeru di 2025

Es Sintinela tene asuntus ku na studadu na 3 di Marsu te 6 di Abril di 2025.



STUDU 1

Bo ngaba Jeova

I na studadu na semana di 3 te 9 Marsu di 2025.



STUDU 2

Kuma ku omis pudi mostra amor ku rispitu pa se minjer

I na studadu na semana di 10 te 16 di Marsu di 2025.

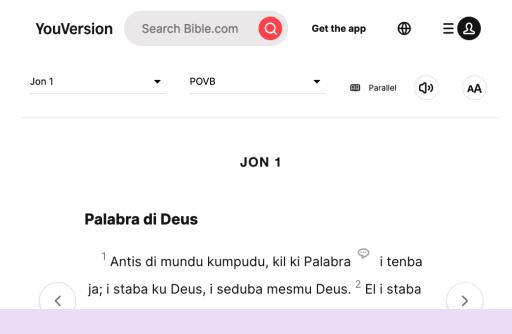


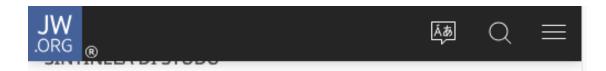
STUDU 3

Toma disisons ku na kontenta Jeova

I na studadu na semana di 17 te 23 di Marsu di 2025.

Machine translation for Guinea-Bissau Creole (Kiriol)





Janeru di 2025

Es Sintinela tene asuntus ku na studadu na 3 di Marsu te 6 di Abril di 2025.



STUDU 1

Bo ngaba Jeova

I na studadu na semana di 3 te 9 Marsu di 2025.

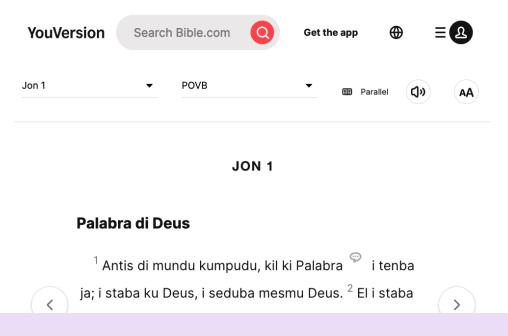


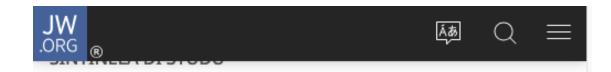
STUDU 2

Kuma ku omis pudi mostra amor ku rispitu pa se minjer

I na studadu na semana di 10 te 16 di Marsu di 2025.

Machine translation for Guinea-Bissau Creole (Kiriol)





Janeru di 2025

Es Sintinela tene asuntus ku na studadu na 3 di Marsu te 6 di Abril di 2025.



STUDU 1

Bo ngaba Jeova

I na studadu na semana di 3 te 9 Marsu di 2025.



STUDU 2

Kuma ku omis pudi mostra amor ku rispitu pa se minjer

I na studadu na semana di 10 te 16 di Marsu di 2025.

Source	Domain	# Sentences
Bible	Religious	29,876
(Old Testament)		(22,220)
(New Testament)		(7,656)
JW WT series	Semi-Religious	6,880
JW Donations series	Semi-Religious	219
Bilingual dictionary	General	1,603
All		38,578

Table 1: Number of sentences collected from each data source. This does not include the 1,983 lexical items also collected from the dictionary.

WT = Watchtower (a Jehovahs Witnesses monthly publication)

Source	Domain	# Sentences
Bible	Religious	29,876
(Old Testament)		(22,220)
(New Testament)		(7,656)
JW WT series	Semi-Religious	6,880
JW Donations series	Semi-Religious	219
Bilingual dictionary	General	1,603
All		38,578

WT = Watchtower (a Jehovahs Witnesses monthly publication)

Table 1: Number of sentences collected from each data source. This does not include the 1,983 lexical items also collected from the dictionary.

Figure 1: Average performance of Portuguese-Kiriol, Kiriol-Portuguese, Kiriol-English and English-Kiriol models trained on different portions of Bible and Watchtower data when used to translate a test set of 1,000 domain-general dictionary sentences. Standard errors across model sets shown with error bars.

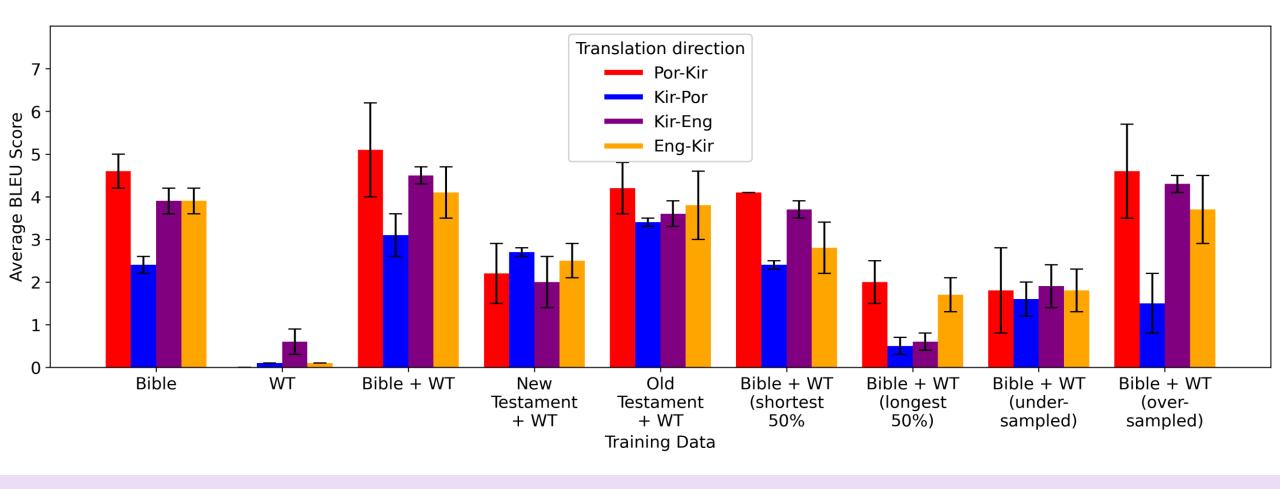


Figure 1: Average performance of Portuguese-Kiriol, Kiriol-Portuguese, Kiriol-English and English-Kiriol models trained on different portions of Bible and Watchtower data when used to translate a test set of 1,000 domain-general dictionary sentences. Standard errors across model sets shown with error bars.

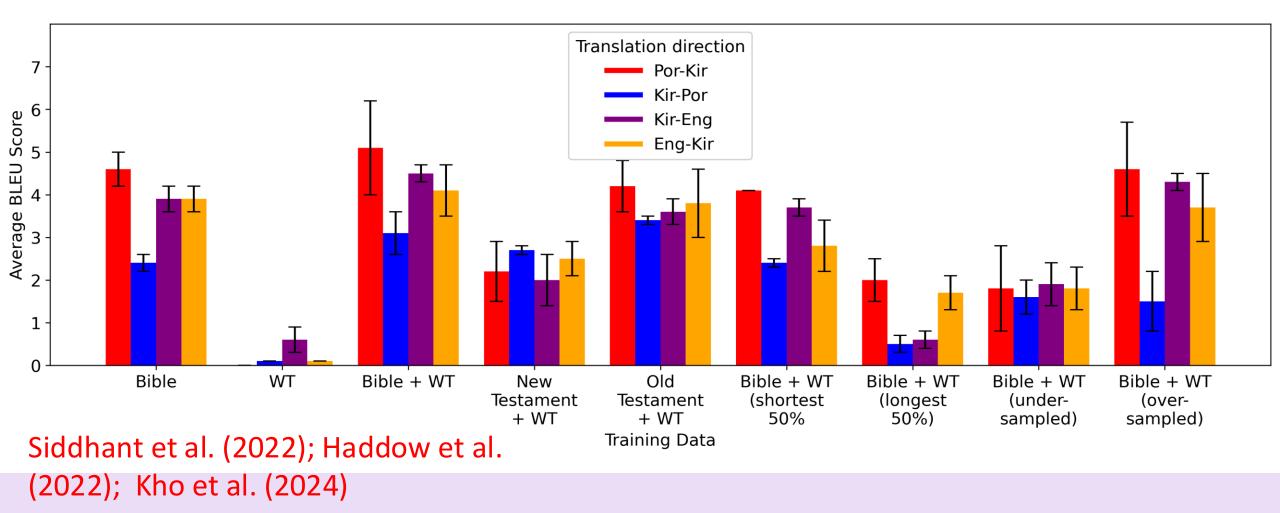


Figure 1: Average performance of Portuguese-Kiriol, Kiriol-Portuguese, Kiriol-English and English-Kiriol models trained on different portions of Bible and Watchtower data when used to translate a test set of 1,000 domain-general dictionary sentences. Standard errors across model sets shown with error bars.

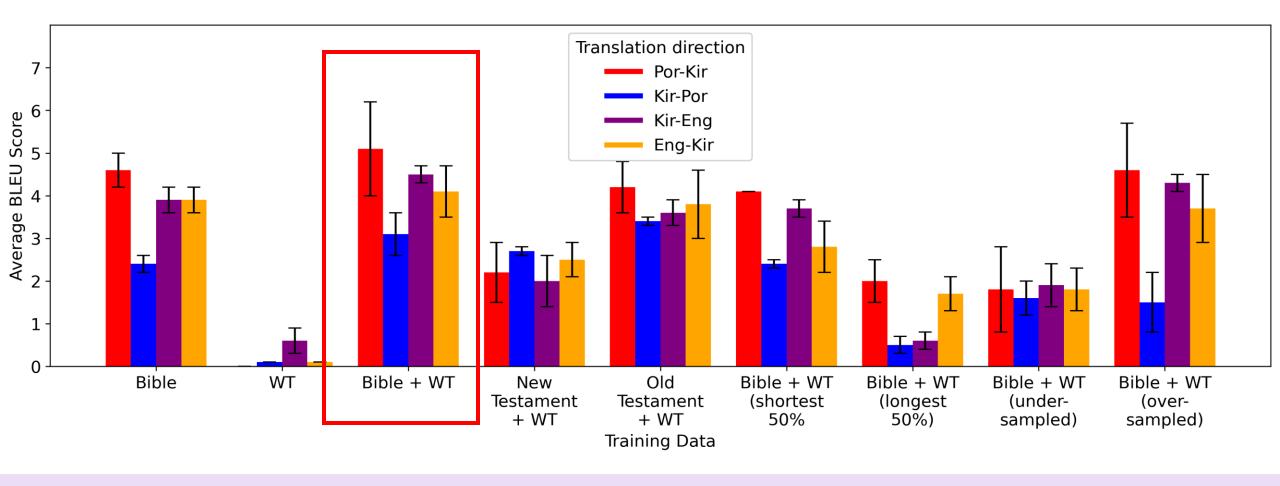


Figure 1: Average performance of Portuguese-Kiriol, Kiriol-Portuguese, Kiriol-English and English-Kiriol models trained on different portions of Bible and Watchtower data when used to translate a test set of 1,000 domain-general dictionary sentences. Standard errors across model sets shown with error bars.

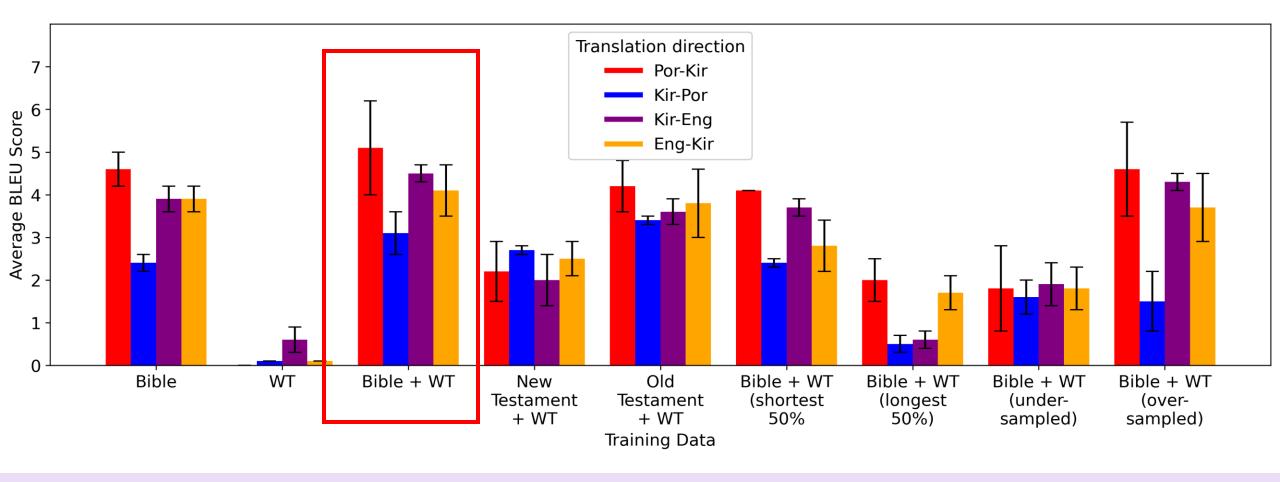
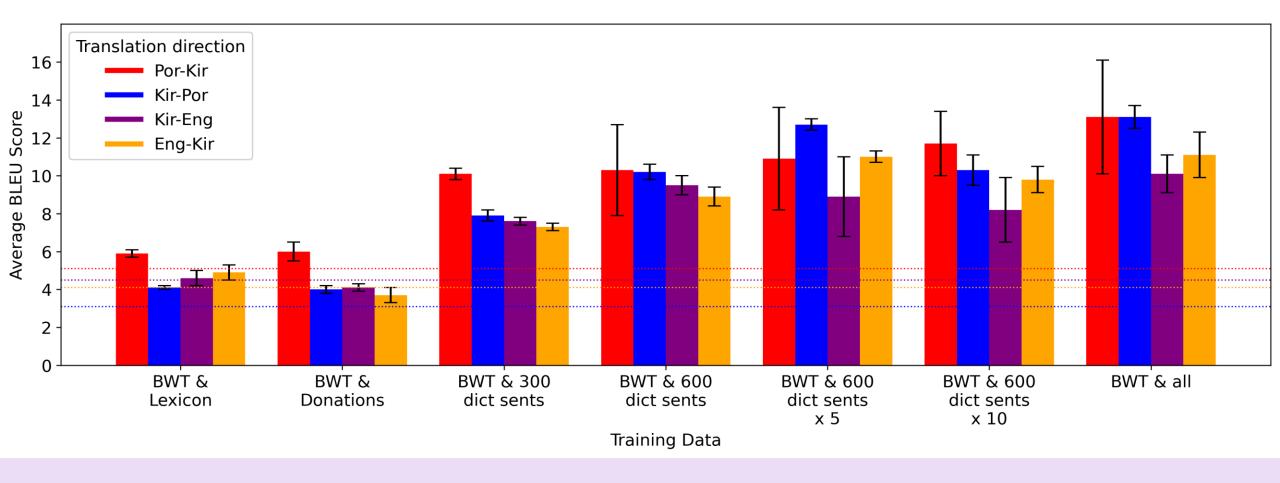


Figure 2: Average performance of Portuguese-Kiriol, Kiriol-Portuguese, Kiriol-English and English-Kiriol models trained on Bible, Watchtower and different combinations of domain-general data when used to translate test set of 1,000 domain-general dictionary sentences. Standard errors across model sets shown with error bars, and **baseline average performance of models** trained only on Bible and WT data is shown with dotted lines.



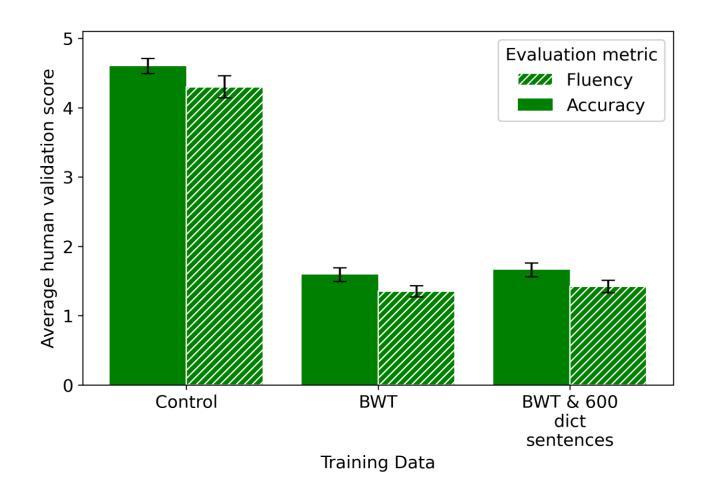


Figure 3: Average scores across all language directions of human judgements for accuracy (solid) and fluency (hatched) of translated sentences from the reference sets (control) and from models trained on Bible and WT data (BWT) and Bible, WT and 600 dictionary sentences. Standard errors across all judgements for each condition are shown with error bars.

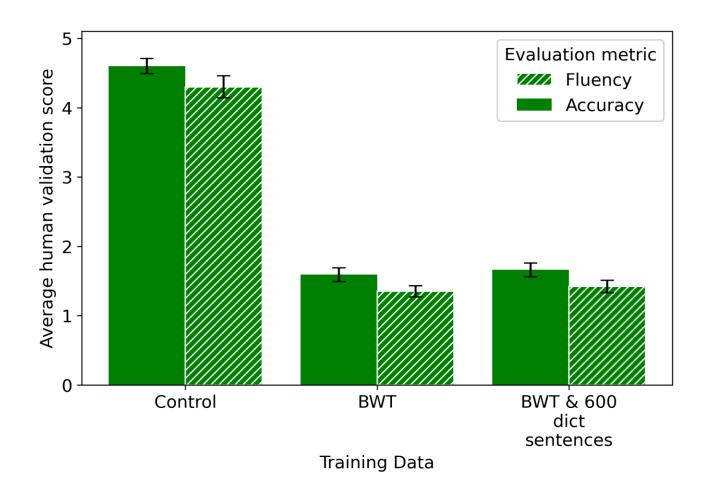


Figure 3: Average scores across all language directions of human judgements for accuracy (solid) and fluency (hatched) of translated sentences from the reference sets (control) and from models trained on Bible and WT data (BWT) and Bible, WT and 600 dictionary sentences. Standard errors across all judgements for each condition are shown with error bars.

- the increases in BLEU scores were still too low overall to be perceptible to the human eye
- the overall utility of the models trained on mostly religious data remains limited

Bianda sufisienti pa tudu djintis. There's enough food for everyone.

Give me the word to all the people.

N'baiba bisita ___ ña primu I was going to visit my cousin.

I went to the Mishonite.

- the increases in BLEU scores were still too low overall to be perceptible to the human eye
- the overall utility of the models trained on mostly religious data remains limited

Para mima bu ermon.

Stop spoiling your brother.

Paradise, be your brother.



Yes!

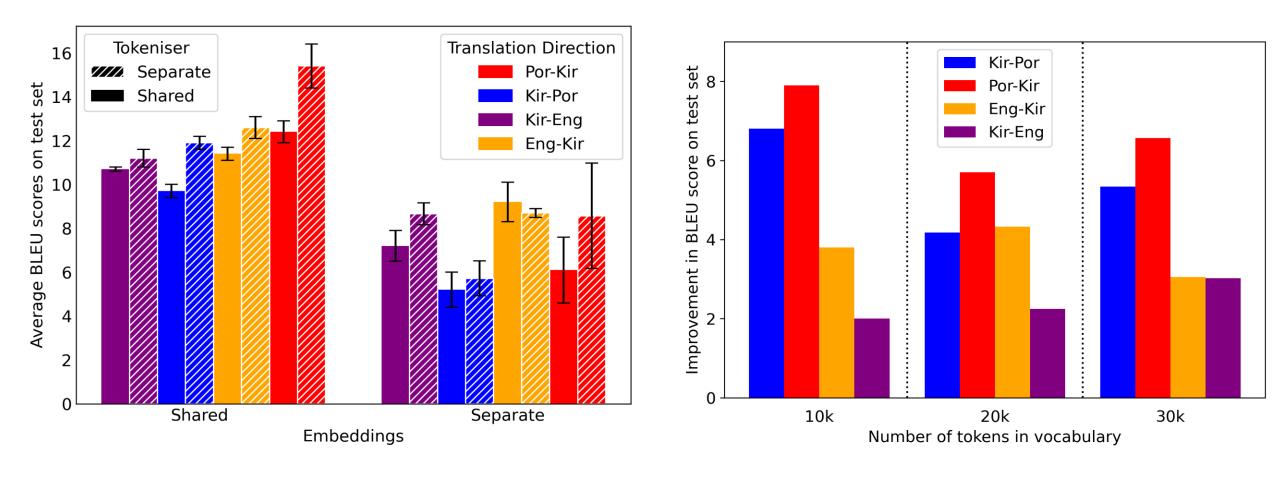
1. Kiriol-Portuguese tokenisers have more overlapping vocabulary items than Kiriol-English tokenisers

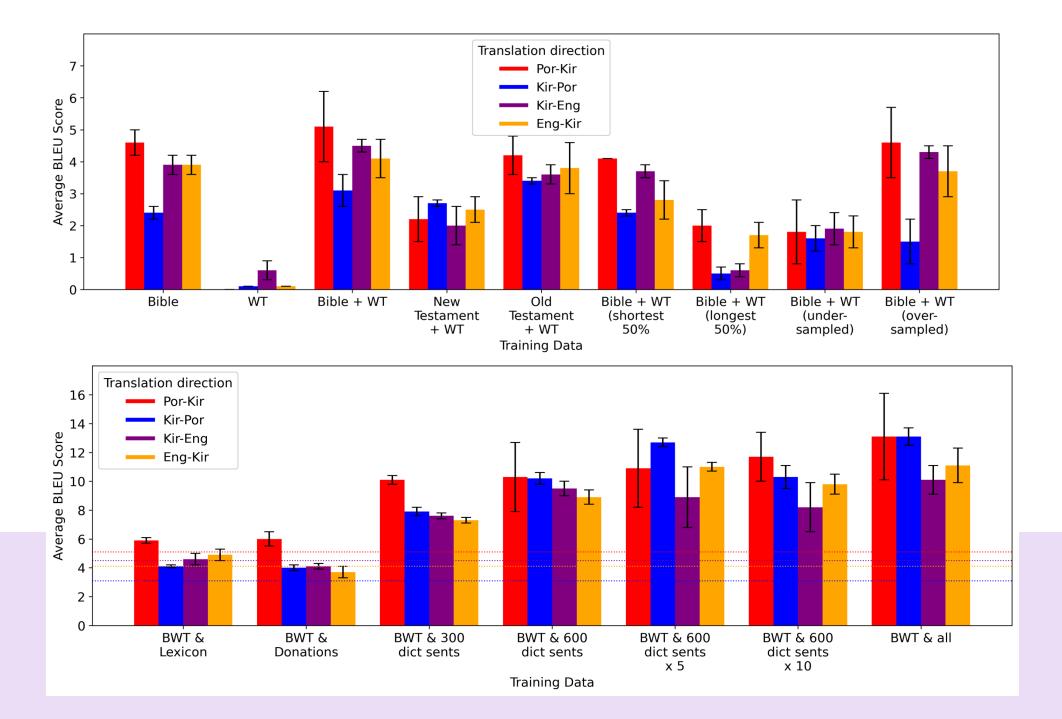
Yes!

- 1. Kiriol-Portuguese tokenisers have more overlapping vocabulary items than Kiriol-English tokenisers
- 2. Compared to using shared tokenisers, using separate tokenisers worsens fertility on Portuguese texts more than on English texts

Yes!

- Kiriol-Portuguese tokenisers have more overlapping vocabulary items than Kiriol-English tokenisers
- 2. Compared to using shared tokenisers, using separate tokenisers worsens fertility on Portuguese texts more than on English texts
- 3. Compared to using shared embeddings, using separate embeddings reduces performance for Kiriol-Portuguese and Portuguese-Kiriol more than Kiriol-English and English-Kiriol





Adding small amounts of target domain data to religious training datasets considerably increase BLEU scores on the target domain...

Adding small amounts of target domain data to religious training datasets considerably increase BLEU scores on the target domain...

...but these are not reflected in human judgements as BLEU scores remain low overall.

Adding small amounts of target domain data to religious training datasets considerably increase BLEU scores on the target domain...

...but these are not reflected in human judgements as BLEU scores remain low overall.

Shared vocabulary between Kiriol and Portuguese does improve tokenization efficiency and model training...

Adding small amounts of target domain data to religious training datasets considerably increase BLEU scores on the target domain...

...but these are not reflected in human judgements as BLEU scores remain low overall.

Shared vocabulary between Kiriol and Portuguese does improve tokenization efficiency and model training...

...but these effects are somewhat complicated by the differences in morphological complexity of the languages involved.











What do you think would be the next best steps for this work?

What other methods should we look at for investigating impact of lexical overlap on creole-lexifier MT?

What role for participation by creole communities?

Thank you!







Paper



Me