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GuyLingo: The Republic of Guyana Creole Corpora

Anonymous NAACL submission

Abstract

While major languages often enjoy substantial attention and resources, the linguistic diversity across the globe encompasses a multitude of smaller, indigenous, and regional languages that lack the same level of computational support. One such region is the Caribbean. While commonly labeled as "English speaking," the ex-British Caribbean region consists of a myriad of Creole languages and dialects thriving alongside English. In this paper, we present **GuyLingo**: a comprehensive corpus designed for advancing NLP research in the domain of Creolese, the most widely spoken language in the culturally rich nation of Guyana. We first outline our framework for gathering and digitizing this diverse corpus, inclusive of colloquial expressions, idioms, and regional variations. We then demonstrate, alongside discussions with Creolese experts, the challenges of training and evaluating NLP models for machine translation for Creolese. Lastly, we discuss the unique opportunities presented by recent NLP advancements for accelerating the formal adoption of Creole languages in the Caribbean.

1 Introduction

Major languages such as English and Chinese frequently receive considerable attention and resources due to their global prominence and economic influence (Lent et al., 2021, 2022a). The extensive focus on these major languages in natural language processing (NLP) has resulted in the development of sophisticated models, extensive datasets, and digital applications consumed by millions of users today. However, despite this global prominence, the linguistic landscape of the globe extends far beyond these dominant languages, encompassing a plethora of smaller, indigenous, and regional languages that play crucial roles in the cultural heritage and communication of their respective communities (Lent et al., 2022c; Hersh-



Figure 1: Map of Guyana and its neighboring territories

covich et al., 2022). The Caribbean Community (CARICOM) is an example of one such region.

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Within the diverse linguistic tapestry of the Caribbean Community, a rich array of languages thrives, reflecting the historical, cultural, and ethnic diversity of the region (Rickford, 1987; Holbrook and Holbrook, 2001). While English is commonly used as the official language in many CARICOM member states, the linguistic heritage goes beyond just English, encompassing a variety of Creole languages, indigenous languages, and influences from African, Indigenous, European, and Asian languages (Devonish and Thompson, 2013).

Creole languages, often born out of the historical context of slavery and colonialism, hold a significant place in the linguistic mosaic of the Caribbean. These languages, such as Jamaican Patois (Armstrong et al., 2022), Trinidadian Creole (Michaelis et al., 2013), and Haitian Creole (Hewavitharana et al., 2011), have evolved as vibrant means of communication, blending elements of African languages with those of European colonizers (Hagemeijer et al., 2014b).

Creole	English
"It luk laik nof ting cheenj op," Seera se. "Somtaim mi doz get	"So many things feel like they have changed," said Sara. "I get
fraikn."	scared about it sometimes."
When me lef' han' 'cratch me, money a-come	When my left hand itches, money is coming.
Di leedii prapa nais	The lady is very pretty

Table 1: Example Guyanese Creole from GuyLingo and its English Translation

Despite its prominence as the mother tongue of the majority of the over 700,000 inhabitants of the Republic of Guyana, Creolese is often subject to challenges in terms of recognition and preservation due to its co-existence with British English, the sole official language of the country. Like other Creole languages, this is partially due to the deeply rooted social attitudes and perceptions about language prestige, resulting in limited efforts to digitize, document, and promote Creolese in official and educational contexts (Hershcovich et al., 2022).

In this work, we introduce GuyLingo, a corpus for Guyanese Creole curated for advancing NLP research and development in Creole. Using these sources, we explore the machine translation task between English and Creolese. To aid in this process we design and implement the Guyanese Creole Translation tool¹, a web-based GPT-powered machine translation tool. Lastly, we briefly discuss insights gained from these developments for accelerating the formal adoption of Creole languages in the nation and potentially the Caribbean Region.

2 GuyLingo Corpus

This section describes the curation of GuyLingo, a corpus of Creolese, the primary spoken language of Guyana. The creation of this corpus aims to address the scarcity of resources and attention devoted to indigenous and regional languages within the NLP community.

2.1 Data Collection

The compilation of GuyLingo involves the collecting and digitizing of a series of linguistic resources encompassing a spectrum of Creolese expressions, idiomatic phrases, and regional variations. To ensure inclusivity and authenticity, we employ a multi-pronged approach:

2.1.1 Expert Collaboration

In collaboration with [removed for anonymity], a collection of original Guyanese Creole sources were curated, digitized, and manually transcribed by a team of researchers. Examples of this include Peirs (1902) a book of Guyanese proverbs, containing over 1k culturally rich proverbs from early British Guiana times still used today, and (Helen Patuck, 2020) a COVID-19 children's book transcribed by Creolese experts for primary education students. In addition, our team of native Creole experts manually construct a corpus of high-quality common Guyanese Creole sayings and terms. Table 2 shows a full breakdown of all information sources.

2.1.2 Online Resources

While no nationally adopted writing system for Creolese exists, many web-based sources, such as language forums, blogs, educational platforms, etc., contain small excerpts of colloquialisms, every-day conversations, and idiomatic expressions prevalent in the Guyanese Creole. These sources were scraped, cleaned, verified, and added to GuyLingo as shown in 2.

2.2 Dataset Characteristics

GuyLingo encapsulates a diverse array of linguistic data, including but not limited to:

- Conversational dialogues
- Idiomatic expressions and phrases
- Proverbs and folklore
- Regional variations and dialectical nuances

In total, GuyLingo consists of 1969 Guyanese Creole sentences with a vocabulary size of 4177 unique Creole words.

3 GuyLingo for Machine Translation

To investigate the utility of GuyLingo, we conduct experiments in the setting of machine translation to assess the ability of NLP models to facilitate English←→Guyanese Creole translation. As such to enable the training and evaluation of these models GuyLingo was further expanded to include English Creole translation pairs. Of the 1969 sentences, the Common Guyanese Creole Saying

¹https://github.com

Sources	Type	# Sentences	Vocab Size
Guyanese-Creole-English Vocabulary-Basic words.	Corpus	20	71
(Polyglot Club, Accessed 2023)			
Guyanese Creole. (Wikipedia, Accessed 2023)	Article	6	28
Gender and Pronominal Variation in an Indo-Guyanese	Journal Article	20	82
Creole-Speaking (Sidnell, 1999)			
Review of Guyanese Creole English (Guy, Accessed	Presentation	9	96
2023)			
Guyanese Creole Survey Report. (Holbrook and Hol-	Language Survey	7	45
brook, 2001)			
APiCS Online -Structure dataset. (Michaelis et al., 2013)	Report	176	351
Creolese. (Devonish and Thompson, 2013)	Journal Article	49	112
Habitual and Imperfective in Guyanese Creole. (Sidnell,	Journal Article	50	103
2002)			
Tense and aspect in Guyanese Creole: A syntactic, se-	PhD Thesis	191	374
mantic and pragmatic analysis (Gibson, 1982)			
Two areas of Guyanese Grammar (Guyanese Languages	Article	14	26
Unit, 2016)			
Me Na Able: Creolese 101 (Letters from Guyana, 2017)	Blog	8	25
Travel Phrases - Guyanese Creole (Travel Phrases)	Blog	4	9
My Hero is you (Helen Patuck, 2020)	Educational	114	831
The Proverbs of British Guiana (Peirs, 1902)	Book	1060	2054
Common Guyanese Creole Sayings (Manually created by	Corpus	302	712
experts)			
Total		1969	4177

Table 2: Compilation of Guyanese Creole Language Resources: Sources, Type, Sentences, and Vocabulary Size

corpus was manually transcribed into English. In addition, 339 common Creole terms from (Peirs, 1902) alongside their English pairs we extracted and verified. Using these initial translation pairs, the Guyanese Creole Translation Tool was built to allow the initial translation of remaining sentences in GuyLingo.

3.1 Guyanese Creole Translation Tool

The Guyanese Creole Translation tool, as shown in figure 4, is a web-based application built using Django+React to facilitate easily storing, editing, and iterative testing of English Creole translations. The UI allows Creolese experts to easily enter text in English or Creole and get a sample translation. We utilize GPT-4 (OpenAI, 2023) to automatically perform these translations. The advanced prompt includes a subset of example verified translations from GuyLingo as in-context examples for generation. Once prompted, the user can modify the generated output before saving it to the database. Users also have the option to modify the advanced prompts as well as provide more seed examples for greater control over the translation process. For instance, users can provide a Guyanese proverb and instruct GPT4 to consider the nuances of the Guyanese culture while translating the text. The remaining translation pairs from GuyLingo were generated and verified using this tool, resulting in 1969 total translation pairs. For training and evaluation, we use our 302 manually curated translation pairs for testing and the remaining translation pairs for model training.

3.2 Experiment Setup

Training and Models We consider the models of T5 (Raffel et al., 2023), BART (Lewis et al., 2020) and Pegasus (Zhang et al., 2020) for their demonstrated performance on several machine translation tasks. All models were implemented with PyTorch and Hugging Face Transformers. We train all models with AdamW (Loshchilov and Hutter, 2019) and a weight decay of 0.01. We use a learning rate of 2e-5, batch size of 4, and a linear learning rate warmup over the first 10% steps with a cosine schedule. We pre-process the data and train all models with varying random seeds over multiple runs for 10 epochs. Approximately 200 GPU hours were required to train all hyperparameter variations across all tasks.

Evaluation For automatic evaluation metrics, we adopted the common methods used for language generation based on n-gram overlap: BLEU (Papineni et al., 2002), ROUGE (Lin, 2004) and METEOR (Banerjee and Lavie, 2005).

3.3 Results

Table 3 and 4 summarize our evaluation results on automated metrics. For en-creole translation, our

Model	Bleu	Rouge1	Rouge2	RougeL	Meteor
T5-Large	09.74	37.44	13.74	36.63	28.19
MBart-Large	12.02	38.39	17.02	37.58	31.33
Bart-Large	12.11	40.56	18.47	39.64	32.77
Bart-Base	10.17	37.49	16.08	36.59	29.54
Pegasus-Large	02.67	24.15	05.30	23.16	16.38

Table 3: Performance of MT Models on English-Creole Translation

Model	Bleu	Rouge1	Rouge2	RougeL	Meteor
T5-Large	19.70	47.71	26.99	46.47	42.45
Bart-Large	17.70	45.74	24.41	39.75	32.77
Bart-Base	14.20	41.68	20.04	40.40	35.95
Pegasus-Large	6.10	28.96	09.88	27.91	22.53

Table 4: Performance of MT Models on Creole-English Translation

results show that the Bart-Large model achieves the best performance amongst all models with a BLEU score of 12.11, ROUGE-1 score of 0.41, ROUGE-2 score of 0.18, ROUGE-L score of 0.40, and METEOR score of 0.33. whereas on creoleeng translation, T5-Large achieves the best performance with a BLEU score of 19.70, ROUGE-1 score of 0.48, ROUGE-2 score of 0.27, ROUGE-L score of 0.46, and METEOR score of 0.42. From manual inspection of both settings, we find that the models struggle mainly with en-creole translation due to incoherent mapping of certain key creole terms to English.

4 Discussion

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In this section, we briefly discuss the unique opportunity presented by recent NLP advancements for accelerating the formal adoption of Creole languages in the Caribbean.

4.1 AI-Driven Applications for Native Languages

One of the major issues affecting the formal adoption is Creolese despite its prominence as a spoken language is its lack of use in formal communication outlets such as literature, news, and written texts. AI-driven applications fueled by rich data sources such as GuyLingo present a major opportunity for enabling the development of educational content, legal documents, and official communications in Creolese. Figure 2 showcases a conversational AI Assistant named [removed for anonymity] deployed to citizens of Guyana speaking in Creolese fueled by GuyLingo. Such applications present the ability to make Creolese more accessible and applicable in various formal contexts further allowing

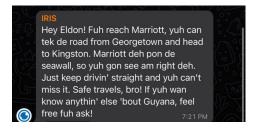


Figure 2: Conversational Agent in Whatsapp speaking in Guyanese Creole.

citizens to feel empowered and foster a sense of national pride.

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5 Related Works

In the context of linguistic diversity, prior works (Hershcovich et al., 2022; Lent et al., 2021, 2022a) have highlighted the challenges faced by lesserknown languages, emphasizing the importance of recognition and preservation. Works such as Dabre and Sukhoo (2022), Hagemeijer et al. (2014a), and Liu et al. (2022) have contributed to advancing NLP research in Creole languages by building a corpus of text for various Creole languages, fostering machine translation, and enhancing language modeling techniques specific to these linguistic varieties. Our work falls into this category. On the other hand, works such as Lent et al. (2022b) and Lent et al. (2022c) emphasize the importance of linguistic diversity by documenting the challenges and exploring the complexities of language modeling for underrepresented languages. The juxtaposition of these studies with the dominance of major languages in NLP underscores the need for more inclusive research efforts that consider the linguistic richness and cultural significance of smaller, indigenous languages within global technological advancements.

6 Conclusion

In this paper, we introduce GuyLingo, a corpus of Guyanese Creolese designed to facilitate advancements in NLP research. We discuss the process of gathering and digitizing this diverse corpus while highlighting the unique opportunities presented by recent NLP advancements for accelerating the formal adoption of Creole languages in the Caribbean. By providing access to a rich collection of colloquial language expressions, idioms, and regional variations, we hope to encourage further research in this field and improve the representation and understanding of Creole languages in NLP.

7 Limitations

While our work aims to contribute to the advancement of NLP for Creole, several limitations arise:

Limited Representation: Guyana is home to many languages outside of Creolese such as Wapichan, Makushi, Wai Wai, Akawaio, Arekuna, Patamuna, Kalina (Carib), Warrau, and Lokono to name a few. Given the cultural significance of these languages, future research should prioritize their inclusion to ensure a more inclusive and representative dataset. Additionally, The rich tapestry of languages in the region extends beyond Guyanese Creole, and efforts should be made to include additional Creole languages and dialects for a more comprehensive understanding.

Limited Generalizability: The findings and insights gained from our work, particularly regarding the formal adoption of Creole languages, may have limited generalizability to other regions or linguistic contexts.

Language Evolution: Creole languages, by their nature, are dynamic and subject to continuous evolution. The static nature of a curated corpus and machine translation models may not fully capture the evolving linguistic landscape, necessitating regular updates and adaptations to reflect current linguistic usage.

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		Creole Translation tool detailing our prompts and	454
103	Zhengyuan Liu, Shikang Ni, Aiti Aw, and Nancy Chen.		
104	2022. Singlish message paraphrasing: A joint task	user interface.	455
105	of creole translation and text normalization. In Pro-	Translate the following Guyanese Creole tex	t
106	ceedings of the 29th International Conference on	and provide the resulting English translation	
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804	Ilya Loshchilov and Frank Hutter. 2019. Decoupled	accurate. Guyanese Creole is spoken in Guyar	าล
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111	Susanne Maria Michaelis, Philippe Maurer, Martin	making it comprehensible in English.	
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113	APiCS Online. Max Planck Institute for Evolutionary		
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118	ation of machine translation. pages 311–318.	Creole: 'Taggah	
119	James Peirs. 1902. The Proverbs of British Guiana.	English: Stop-off	
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Figure 3: Example GPT-4 Prompt with translation examples

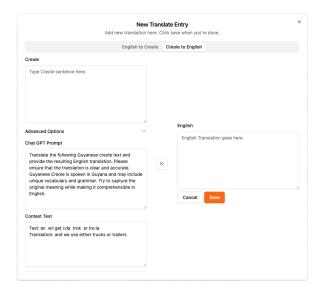


Figure 4: User Interface of Guyanese Creole Translation Tool. This tool allows experts to rapidly and iteratively create translation pairs using GPT-4 (OpenAI, 2023) as a generator.