HAMBAEON: BUILDING A COMPREHENSIVE AKEANON SPEECH AND TEXT CORPUS TOWARDS LANGUAGE PRESERVATION AND DIGITAL INCLUSION

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Abstract

Akeanon, a language spoken in Aklan, Philippines, is classified as a low-resource language (LRL) due to its limited linguistic resources and lack of digital integration. This research aims to develop and establish a comprehensive text and speech corpus, and build a model as a foundation for an automatic speech recognition (ASR) system for standardized Akeanon language. For the methodology, data collection will include compiling word lists for Akeanon based on the Swadesh 207 list and extracted text from an existing Akeanon-to-English dictionary. After compiling the word lists, their phonetic transcriptions will be encoded for the text corpus. For the development of the speech corpus, a selection of 1000 Akeanon words, having words from Swadesh list as priority, will be voice recorded from a total of fifty native speakers with varying gender and age. Audio resources containing the Akeanon language as well as their corresponding transcriptions that are available online will also be retrieved. The development of the speech and text corpus will be overseen and validated by a linguistics expert. A DNN model will be built and trained with Kaldi, using the newly developed corpus for initial results. This study aims to contribute to the preservation and digital inclusiveness of the Akeanon language and lay the groundwork for future works in developing an ASR system for the language.

Keywords: Language resources, Natural language processing (NLP),

Speech recognition, Philippine languages, Aklan, Akeanon,

Language corpus, Low-resource languages (LRL)

Contents

1	Intr	roduction	1
	1.1	Overview	1
	1.2	Problem Statement	3
	1.3	Research Objectives	5
		1.3.1 General Objective	5
		1.3.2 Specific Objectives	5
	1.4	Scope and Limitations of the Research	6
	1.5	Significance of the Research	6
2	Rev	riew of Related Literature	8
	2.1	Automatic Speech Recognition	8
	2.2	Lexicon Model	9

	2.3	Acous	tic Model	10
	2.4	Langu	age Model	10
	2.5		Dialects and Low-Resource Languages On Automatic Speech	
		Recog	nition	11
	2.6	The K	Kaldi ASR Toolkit	12
	2.7	The A	keanon Language	12
		2.7.1	History and its Speakers	12
		2.7.2	Phonology	13
		2.7.3	Morphology	15
		2.7.4	The 300 Languages Project: A Worldwide Linguistic Initiative	17
3	Res	earch	Methodology	20
	3.1	RESE	ARCH ACTIVITIES	20
		3.1.1	Data Collection	21
		3.1.2	Preprocessing	23
		3.1.3	Speech and Text Corpus Development	24
		914		24
		3.1.4	Validation	24

	3.2 Calendar of Activities	25
4	Preliminary Results	27
	4.1 Text and Phonetic Transcription	27
Re	eferences	29
\mathbf{A}	Research Ethics Documents	33
В	Resource Persons	37
\mathbf{C}	Initial Results of the Text Corpus	38

List of Figures

3.1	Research Methodology	20
4.1	Preliminary transcription of Akeanon words into text along with their phonetic representations	28
4.2	Akeanon Dictionary transcription	28
A.1	Screenshot of email from Dr. Zorc granting permission to use the Aklanon-to-English Dictionary.	33
A.2	Screenshot of email from Bible.com granting permission to use audio resources and transcription from their website	34
A.3	Informed consent to be given to the participants of the study	35
A.4	Information sheet attached to the informed consent	36
C.1	Partial results of transcription for the Swadesh 207 Word List	38
C.2	Partial results of transcription of the Aklanon-to-English Dictionary.	39

List of Tables

3.1	Categories of Native Speakers			•	•						23
3 9	Timetable of Activities										26

Chapter 1

Introduction

1.1 Overview

Speech-to-Text (STT) technology has rapidly evolved in recent years, driven by advancements in deep learning algorithms such as recurrent neural networks (RNNs) and convolutional neural networks (CNNs), which have significantly improved the accuracy of STT systems (Televic, 2024). Open-source toolkits such as Kaldi have further accelerated research and development in this field by providing a flexible framework for building and training custom automatic speech recognition (ASR) models. ASR systems, which convert speech into text, have become essential components of various applications, from virtual assistants to transcription services (Cerna et al., 2023). However, despite these advancements, only a few Philippine languages have been explored and integrated into this technology. This special problem focuses on one of the understudied (Wellstood, 2022) Central Philippine languages, Akeanon.

Akeanon is an Austronesian language belonging to the Visayan subgroup (Biray, 2023). With more than 130,000 households (PSA, 2023) speaking the language, Akeanon is primarily spoken in the province of Aklan, located in northwestern Panay. Biray (2023) explains that the language has several dialects, each typically named after the town where it is spoken. These include Akeanon Buruangganon, Akeanon Nabasnon, Akeanon Bukidnon, and the standard Akeanon, which is spoken in Kalibo, the provincial capital of Aklan. For this special problem, the researchers will focus on developing the speech and text corpus specifically for standard Akeanon.

Up to this date, no studies have been conducted that is directly related to Akeanon and speech recognition altogether. However, there exist similar studies in the context of speech recognition on other regional languages such as Bisaya in the study of Cerna et al. (2023), Hiligaynon, studied by Billones and Dadios (2014) and Panizales et al. (2023), and in the study of Liao et al. (2019) for Bikol and Kapampangan. This special problem aims to bridge the gap in speech recognition for Akeanon starting with establishing a foundational speech corpus for the language, which can lay the groundwork for future research and applications. The corpus development will draw on methodologies from similar studies conducted for other regional languages such as the study of Cerna et al. (2023) and Liao et al. (2019), adapting them to meet the specific needs of Akeanon. In doing so, the project aims to bring Akeanon closer to digital integration, promoting inclusivity in speech recognition technology for Philippine languages. By bridging this gap, this special problem aspires to create a resource that can benefit future ASR developments, language preservation efforts, and the broader field of computational linguistics.

Creating a speech-to-text (STT) system for the Akeanon language not only fills the gap in representation for this regional language but also aids in its preservation and fosters digital inclusion. This specific project aims to establish a foundational corpus that effectively captures the distinct speech patterns and intricacies of Akeanon, while taking into account the language's unique phonetic and linguistic features. Utilizing the resources gathered for this research, the team will concentrate on developing a comprehensive text and speech corpus that can provide a basis for future speech recognition systems pertaining to the Akeanon language. The researchers will also build and train a DNN model using Kaldi toolkit, to develop an ASR system that will provide initial speech recognition results for Akeanon. Finally, the study intends to investigate the challenges faced in developing speech models for languages with limited resources, offering valuable insights for the wider field of speech technology development.

1.2 Problem Statement

Akeanon remains underrepresented in modern speech technologies. According to Khan et al. (2023), in machine learning, natural language can be categorized into two categories: low-resource languages (LRLs) and high-resource languages (HRLs). Among these resources are (a) collections of text in different formats, such as research papers, journal articles, social media content, etc.; (b) lexical, syntactic, and semantic resources, such as dictionaries, bag of words, semantic databases, etc.; and (c) task-specific resources, such as annotated text, machine translation corpora, part-of-speech tags, etc.. HRLs e.g. English, French, Japanese, etc., are languages that are highly accessible and have many data re-

sources that can be used for natural language processing (NLP). LRLs, on the other hand, are understudied and have few data resources that can be utilized for NLP. Most regional languages in the Philippines are considered to be LRL, including the Akeanon language. Alejan et al. (2021) raised concerns on the Philippines' inclusion on a global list of the top ten "language hotspots", which means that many of its languages are disappearing faster than they are being completely documented. Their study noted the global rate of language extinction, which is one in every two weeks. They also projected that around half of the 6,000 languages will become extinct by the end of the century, to which most of them are indigenous languages. According to Magueresse et al. (2020) [29], a language supported by NLP techniques can help preserve it from extinction. It will also make the language more available and accessible in digital format, which offers significant commercial value, societal purpose, and applications in a variety of domains (Tsvetkov, 2017).

This special problem aims to address the lack of resources, availability, and accessibility of the Akeanon language in, but not limited to, modern speech technologies by building and establishing a speech and text and corpus for the language. Additionally, by developing an ASR model that is specific for Akeanon would lay the foundation for future research in speech-to-text, and other modern speech technologies for the language. Lastly, this special problem seeks to inspire innovation and drive similar efforts to preserve and develop accessible language technologies for other regional languages in the Philippines.

1.3 Research Objectives

1.3.1 General Objective

The general objective of this study is to construct and establish a comprehensive text and speech corpus for the Akeanon language, which can serve as a foundation for future development of language technologies and automatic speech recognition (ASR) systems. Additionally, the study aims to design and implement an ASR system for the language using the Kaldi toolkit.

1.3.2 Specific Objectives

Specifically, the study targets to:

- 1. Develop an Akeanon text corpus by collecting existing language resources such as dictionaries, word lists, and phonetic transcriptions, and organizing them into an annotated dataset.
- 2. Build a speech corpus by recording native speakers and using pre-existing Akeanon audio resources which can be found online.
- 3. Validate the text and speech corpus with the assistance of linguistic experts and native speakers to ensure accuracy and reliability.
- 4. Develop and evaluate an automatic speech recognition (ASR) model using deep neural networks (DNN) and the Kaldi toolkit with the newly created Akeanon corpora.

1.4 Scope and Limitations of the Research

The system is specific to the standard Akeanon language that is spoken in Kalibo, Aklan, the provincial capital. It is limited to the standard Akeanon even though the language has several dialects, due to resource constraints and the need for specificity. The study is centered around gathering audio samples from native speakers of standard Akeanon to guarantee precision and uniformity in the training of the model. This does not include other variations or dialects that are based on standard Akeanon language such as Buruanganon, Akeanon Nabasnon, and Akeanon Bukidnon, which can have different and unique phonetic and lexical traits. Nevertheless, the model's effectiveness might be influenced by the scarce availability of Akeanon data, potentially affecting its wide-ranging applicability.

1.5 Significance of the Research

Akeanon language, like many indigenous languages in the Philippines, lacks representation in digital technologies. Creating an automatic speech recognition (ASR) model for Akeanon language will help contribute to the preservation of the language in digital format, establishing a resource that will support documentation and education initiatives in the future. The dataset and model produced in the study of Akeanon language can act as a basis for further and additional linguistic research.

Akeanon and its incorporation in speech recognition technology fosters digital inclusivity. This enables Akeanon speakers to engage with technology in their

mother tongue highlighting the areas in education, communication, and public service where language barriers are almost present when accessing the said areas. Mobile applications, virtual assistants and other tools can embed speech to text of Akeanon language to help enhance accessibility and boost engagement with technology.

The challenge faced and lessons learned from this study will help contribute to addressing the lack of representation of low-resource language in AI technology, aligning with the need for inclusivity in language processing (Poupard, 2024). This initiative will help in promoting linguistic diversity as well as safeguard cultural heritage through Akeanon speech recognition in technological advancement. Poupard (2024) highlights that even minimal focus on languages with fewer resources can significantly influence their viability in an increasingly digital world where larger languages prevail.

Chapter 2

Review of Related Literature

2.1 Automatic Speech Recognition

Automatic Speech Recognition (ASR) is a technology that processes human speech into readable text by the use of machine learning or artificial intelligence (AI). The ASR system has grown popular over the past decade as it quickly approaches human accuracy levels, there is a great demand for applications taking advantage of ASR technology in their products to make audio and video data more accessible (Foster, 2023).

Automatic Speech Recognition independently decodes and transcribes spoken language using a machine-base process. An ASR system takes in acoustic signals from a speaker via a microphone, analyzes these signals using various patterns, models, or algorithms, and generates an output, most commonly in text form (Levis & Suvorov, 2012). The importance of differentiating speech recognition

from speech understanding (speech identification) is that, speech understanding focuses on interpreting the meaning of an utterance rather than merely transcribing it. Furthermore, speech recognition is distinct from voice recognition: speech recognition pertains to a machine's capability to identify the words spoken, while voice recognition relates to a machine's ability to discern the manner of speaking (Levis & Suvorov, 2012).

2.2 Lexicon Model

The lexicon model is essential in automatic speech recognition, serving as the bridge between the acoustic representation and the sequence of words produced by the speech recognizer. The lexicon's function can be viewed in two aspects: it first identifies the words or lexical items recognized by the system, and second, it offers the framework to develop acoustic models for each entry (Adda-Decker & Lamel, 2000). Consequently, lexical design consists of two primary components: determining and selecting the vocabulary items and representing each pronunciation entry using the fundamental acoustic units of the recognizer. In large vocabulary speech recognition, the vocabulary is typically chosen to optimize lexical coverage within a specified size of the lexicon, and the basic units selected are generally phonemes or phone-like units ((Adda-Decker & Lamel, 2000).

2.3 Acoustic Model

Acoustic modeling is a fundamental and preliminary step in the process of speech recognition. The acoustic model defines the relationship between acoustic data and linguistic elements. Most calculations in acoustic modeling are attributed to feature extraction and statistical representation, making it a crucial factor in the recognition process. Statistical representations are derived from the features that have been extracted (Bhatt et al., 2020). In the acoustic model, the distribution of these extracted features corresponding to specific sounds is modeled to create a connection between the features and the structures of the linguistic units.

According to Bhatt et al. (2020), several techniques for feature extraction, including those based on human perception and the mechanics of voice production, have been documented. Features were derived for acoustic modeling in a speaker-independent recognition context since such systems pose challenges in speech recognition.

2.4 Language Model

Language models are crucial for various daily applications, including correcting grammatical errors, recognizing speech, and summarizing text. Due to the recent advancements in deep learning techniques, conventional n-gram and word embedding language models are being substituted with neural network-based models (Mago & Qudar, 2020).

Large Language Models (LLMs) have recently shown remarkable abilities, en-

compassing tasks like natural language processing (NLP), language translation, text generation, and answering questions. In addition, LLMs play a vital role in computerized language processing, capable of grasping intricate verbal patterns and producing relevant and coherent responses in various contexts. However, the significant advancements in LLMs have led to a surge in research contributions, making it challenging to fully comprehend the overall impact of these developments (Fahad et al., 2024).

2.5 Local Dialects and Low-Resource Languages On Automatic Speech Recognition

Deep learning technologies have evolved from rudimentary systems to advanced models that can fluently comprehend natural language, making remarkable progress in their integration into Automatic Speech Recognition (ASR). Neural networks have become crucial in ASR for capturing temporal dynamics and phonetic differences, enabling wider use in virtual assistants, educational applications, and customer support (Alharbi et al., 2021). Noisy environments where background sounds significantly impair the accuracy and dependability of speech recognition. The considerable challenge for languages with limited resources is the size of the vocabulary. This influences the performance of the model in which larger vocabularies enhance adaptability but demand more data and computational power. ASR systems struggle with dialectal variation, which can impede model accuracy due to differences in pronunciation, a concern for languages such as Akeanon, known for its various dialects (Alharbi et al., 2021).

2.6 The Kaldi ASR Toolkit

The structure of Kaldi, an open-source toolkit available for speech recognition research, is examined. Kaldi offers a speech recognition framework built on finite-state transducers, utilizing the freely accessible OpenFst, along with comprehensive documentation and scripts for constructing entire recognition systems. Povey et al. (2011) characterized Kaldi as a contemporary toolkit for speech recognition. It is built to be flexible and features one of the more permissive licenses, which enhances its accessibility. Numerous research works have utilized Kaldi in their applications.

2.7 The Akeanon Language

2.7.1 History and its Speakers

Zorc (1995) stated that Akeanon serves as the main language in the northwestern area of Panay Island in the central Philippines, boasting over 350,000 speakers. Both the language and its speakers derive their name from the Akean River, which runs through the heart of the province by the same name. The people, culture, and items linked to this river and region are referred to as Aklanon, while the language is known as Inakeanon, incorporating the -in- infix and an accent alteration, or more generally Bisaya, as Aklanons identify themselves as part of the Visayan cultural and linguistic family. Many Aklanons, particularly those in professional fields, have relocated to various major cities in the Philippines, such as Manila, Cebu, and Cotabato, in pursuit of job opportunities, with sizable communities

also found in San Francisco and New York. The dialect discussed here is that of Kalibo, Aklan, the provincial capital and its main commercial hub. Other dialects are linked to the towns of Altavas, Batan, Balete, Banga, Madalag, New Washington, Numancia, Malinao, Lezo, Makato, Tangalan, Nabas, Ibajay, and Libacao—though the latter two show significant divergence, they remain mutually understandable with the others. Two towns exist within Aklan province that feature different dialects—with Buruanga associated with Kinaray-a, and Malay linked to various dialects of Tablas, Romblon. The closest languages to Akeanon are Kinaray-a and Kuyonon, both of which belong to the West Bisayan subgroup of Central Philippine languages.

2.7.2 Phonology

Akeanon Phonology: Historical and Synchronic Perspectives

The Akeanon language, native to the Aklan province in the Philippines, possesses a distinctive phoneme that sets it apart from other Philippine-type languages. Initially recognized as a voiced velar fricative and subsequently categorized as a velar approximant, this phoneme differentiates Akeanon from its linguistic siblings within the Bisayan group, such as Hiligaynon, Cebuano, and Kinaray-a. The initial examinations of this phoneme can be traced back to Scheerer (1920), who analyzed its correlation with Bisayan cognates. Subsequent research by de la Cruz and Zorc (1968) characterized it as a voiced velar fricative, functioning both as a consonant and a semivowel. More recent studies have reiterated its classification as a velar approximant, emphasizing its absence of articulatory turbulence (Zorc, 2005; Rentillo & Pototanon, 2022)

Linguistic Status and Usage of Akeanon

Akeanon is acknowledged as an institutional language according to the Expanded Graded Intergenerational Disruption Scale (EGIDS) and is included in the Mother Tongue-Based Multilingual Education (MTB-MLE) program in primary education. With approximately 500,000 speakers based on recent estimates, the language flourishes in both spoken and written forms, encompassing social media, radio programs, and public signages. Its phonological framework, which is defined by a three-vowel inventory and distinctive consonantal reflexes, has been influenced by historical changes and cross-linguistic interactions.

Crosslinguistic Comparisons and Historical Accounts

The evolution of the Akeanon phoneme is believed to reflect more extensive linguistic trends, such as velarization and palatalization, seen in various languages. Rentillo and Pototanon (2022) contend that the development of the phoneme may have been shaped by regional linguistic changes or historical interactions with other Bisayan dialects. Moreover, historical accounts from figures such as de Méntrida (1841) and Monteclaro (1957) indicate cultural and linguistic connections to Borneo, which influenced the distinct characteristics of Akeanon speech.

Acoustic and Articulatory Characteristics

Recent acoustic studies conducted by Rentillo and Pototanon (2022) offer empirical insights that differentiate the velar approximant from other phonemes. Their research demonstrates that the formant frequencies (F1 and F2) of this phoneme are lower than those of vowels, with variations that depend on adjacent phonological contexts. These findings emphasize the phoneme's unique articulatory

properties, confirming its classification as an approximant rather than a fricative.

Implications for Language Documentation

The distinctive attributes of Akeanon phonology reinforce the significance of documenting endangered and lesser-known languages. The Akeanon phoneme acts as a case study for exploring phonological diversity and innovation within Philippine languages. As noted by Rentillo and Pototanon (2022), further research could yield greater understanding of the historical and sociolinguistic elements that influence such unique linguistic features.

2.7.3 Morphology

Morphology and its Role in Language

Morphology, which examines word structures and their smallest meaningful units, is fundamental to comprehending the formation and development of languages. In various languages, including Akeanon, derivational morphology transforms syntactic roles or introduces novel meanings through methods like affixation, reduplication, subtraction, and internal modification of words. These methods not only redefine lexical meanings but also influence word categories like parts of speech (Biray, 2023).

Linguistic Diversity in the Philippines

The Philippines is distinguished by its extensive linguistic variety, containing over 180 distinct languages, predominantly of Austronesian origin. Akeanon, which has approximately 460,000 speakers, belongs to the Malayo-Polynesian language

family and functions as an official language in the province of Aklan. The language shares lexical similarities with Kinaray-a and Kuyunon, accompanied by notable dialectical variations throughout the area.

Akeanon Dialectical Variations

Akeanon dialects—including Standard Akeanon, Buruangganon, Nabasnon, and Bukidnon—display specific linguistic characteristics. These dialects are shaped by their geographical and cultural backgrounds, resulting in differences in structure, word order, and affixation. For example, reduplication serves as a prominent morphological feature that modifies meanings, whereas circumfixes are frequently utilized for the formation of new words. Dialect-specific phonemic variations, such as replacing "l" with "r" in certain instances, further highlight these distinctions.

Social and Cultural Significance

The Akeanon language mirrors the social traits of its speakers, showcasing values such as hospitality and respect. Expressions of endearment and polite language are prevalent in daily interactions, emphasizing the cultural identity of the community. Despite structural differences, the fundamental meanings of expressions remain uniform across dialects, illustrating the language's strength and flexibility.

Challenges and Preservation Efforts

Like many other languages in the Philippines, Akeanon faces challenges stemming from modernization and the growing impact of technology. Initiatives to safeguard the language include its integration into the Mother Tongue-Based Multilingual Education (MTB-MLE) framework and the creation of orthographies that document its linguistic characteristics. Nonetheless, further support from both local

and national organizations is crucial to maintain and promote the language in the face of the rising influence of global languages.

2.7.4 The 300 Languages Project: A Worldwide Linguistic Initiative

The 300 Languages Project, led by The Rosetta Project and The Long Now Foundation, stands as a groundbreaking effort aimed at creating a universal collection of human languages. This project seeks to gather and digitize parallel text and audio data from the 300 most frequently spoken languages around the globe. This extensive initiative addresses the significant shortage of resources for linguistic research, particularly for lesser-known languages, by utilizing volunteer-submitted public domain texts and recordings, all of which will be made available through The Internet Archive.

Linguistic Variety and Digital Visibility

Among the roughly 7,000 languages spoken worldwide, merely 20-30 languages possess a substantial digital footprint, including English, Spanish, and Mandarin. These languages, in conjunction with the next 270-280 most spoken languages, encompass over 90% of the global populace. In contrast, the remaining 10% communicate in one of the 6,700 minority languages, many of which are at risk of extinction due to inadequate digital and physical documentation. The 300 Languages Project highlights the importance of showcasing these minority languages by establishing a scalable "seed corpus" that begins small but is intended to expand sustainably.

Contributions to Multilingual Research and Technological Advancements

This initiative distinguishes itself by merging linguistic preservation with technological innovation. By assembling a large-scale public domain multilingual parallel corpus, the project enables progress in speech recognition, automated translation, and cross-linguistic studies. The absence of such resources has historically limited research and development to a small number of languages with existing corpora. The project's focus on widely translated texts, such as the Swadesh List, the Universal Declaration of Human Rights, and chapters 1-3 of Genesis, ensures extensive applicability for linguistic research and tech applications.

Volunteer-Driven, Scalable Approach

The project's dependence on volunteer-contributed materials highlights its scalability and cost-efficiency. By establishing a comprehensive protocol for language documentation, this effort lays out a replicable model for documenting additional languages beyond the initial 300. The low-cost, community-focused method reflects earlier successful documentation endeavors like the ancient Rosetta Stone, which facilitated the understanding of Egyptian hieroglyphs through parallel texts.

Significance for Language Conservation

The 300 Languages Project plays a crucial role in preserving linguistic diversity by documenting and archiving minority languages that are on the brink of disappearing. By making multilingual resources publicly accessible, the initiative not only benefits researchers but also bolsters educational and cultural preservation efforts worldwide. Its alignment with the ALLOW initiative at the Language Technolo-

gies Institute further demonstrates a collaborative dedication to advancements in speech and language technologies.

Chapter 3

Research Methodology

This chapter discusses the methodology used to develop the text and speech corpus for the Akeanon language, as well as building, training, and testing a model to generate initial results. The chapter will be divided into two main parts: Research activities and the calendar of activities for this special problem.

3.1 RESEARCH ACTIVITIES

Figure 3.1 shows the general overview of the methodology for the development of an ASR system for the Akeanon language.

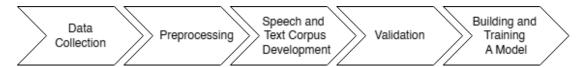


Figure 3.1: Research Methodology

3.1.1 Data Collection

Collating Pre-existing Online Resources

For the data collection, the researchers will make use of existing online resources from the website, Bible.com (see Appendix A for permission to use their resources). These resources include recordings and transcriptions of the Akeanon translations of the multiple books and chapters of the Bible.

Compiling Akeanon Words

The researchers will collect equivalent Akeanon words based off on the Swadesh 207 word-list and use the Aklanon to English Dictionary by Zorc, Reyes, and Prado (1969) as reference. In addition to the Swadesh list, the researchers have reached out to Dr. David Zorc, one of the authors of the Aklanon-to-English Dictionary. Not only who he granted the researchers permission to use the dictionary, but he has also generously provided a soft copy of an existing database containing all the contents of the dictionary to the researchers. This database was shared courtesy of Dr. Jarrette Allen, a linguist who has also worked on the Akeanon language.

Phonetic Transcription

After compiling the Akeanon word lists, their phonetic transcription will also be encoded, using the work of Rentillo and Pototanon (2022) as reference for Akeanon phonology. Audio transcriptions retrieved from the Bible.com will also have their phonetic transcriptions be encoded. For the phonetic transcription the researchers will seek assistance from Ms. Hazel Cipriano, a linguist who is also a native speaker of the language.

Word Selection for Speech Corpus

For building the speech corpus, the researchers will prioritize words from the Swadesh 207 list for the voice recordings. Additionally, to complete the selection of 1000 words that will be prepared for speakers to read, the researchers will collaborate with a linguistic expert during the selection.

Voice Recording

A total of 50 native speakers of standard Akeanon will be gathered for the recording of the generated word list. The 1000 words list will be divided into five sets, with each containing 200 words that are unique to that set. Each set will have 10 speakers for the recording. The researchers have also sought a collaboration with Aklan State University (ASU) - College of Teacher Education for the selection of speakers, with Dr. John Orbista as the primary contact. The speakers will be of varying gender and age. Table 3.1 shows the categories of native speakers. For the audio recording, the microphone that will be used will be Shure SM58 (dynamic, cardiod pick-up pattern) with a Focusrite Scarlett 2i2 audio interface, having Adobe Audition 2021 as the recording software. In case of unavailability, recording over a smartphone with noise-cancelling headphones would suffice. The audio file will be named in the following convention: <gender>_<aspe_group>_<aspe_group>_<aspe_group>_<aspe_group>.<aspe_group>_<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_group>.<aspe_

Ethical Considerations

In the beginning of their session for the voice recordings, participants will be given a consent form containing information relevant to the study. This consent form will serve as a formal acknowledgment of the participant's voluntary involvement

Table 3.1: Categories of Native Speakers

Category	Subcategories
Gender	Male
	Female
	Non-binary
	Others
Age Group	12-15
	16-30
	31-45
	46-65

and understanding of the study's objectives, procedures, and potential risks. The form will clearly explain the purpose of the research, how the data will be used, and the steps taken to ensure confidentiality and anonymity. Participants will be informed that they can withdraw from the study at any time without penalty. Additionally, the consent form will detail the nature of the voice recordings, and the storage of their data. Participants will also be made aware that their voices may be used for research analysis but will not be associated with their personal identities. Upon signing the consent form, participants will be given the opportunity to ask questions and will be reassured that their participation is entirely voluntary and that their rights will be respected throughout the study.

3.1.2 Preprocessing

For preprocessing audio files, Adobe Audition 2021 will be used for recording and audio processing, which will include normalization and noise reduction of the recorded audio. The cleaned up audio files will then be exported in a WAV format.

3.1.3 Speech and Text Corpus Development

The previous steps will set as a precedent for the speech and text corpus development. The development of the text corpus for the Akeanon language involves creating a comprehensive collection of audio transcriptions from pre-existing online resources, dictionary-based word list, and their respective phonetic transcriptions. For the speech corpus, the collated audio transcriptions and word list will be mapped with their corresponding voice recordings and will be annotated accordingly.

3.1.4 Validation

In collaboration with ASU College of Teacher Education and with the guidance of Dr. Zorc, they will be aiding the researchers by validating the newly created text and speech corpora.

3.1.5 Building and Training A Model

To generate initial results for an ASR system, the researchers will build, train, and test a model using the Kaldi toolkit. Similar to the methods employed by Panizales et al. (2023), a ten-fold cross-validation scheme will be used for training and testing the data, with eight folds used for training and two folds used for validation. Finally, the researchers have decided to use a DNN (Deep Neural

Network) model for training.

3.2 Calendar of Activities

Table 3.2 shows a Gantt chart of the activities. Each bullet represents approximately one week worth of activity.

Table 3.2: Timetable of Activities

	Sep	Oct	Nov	Dec	Jan	Feb	Mar	Apr	May
Brainstorming and Selection	•	••						_	
of Topic									
Drafting and Finalization of		••	•						
Chapter 1 - Introduction									
Drafting and Finalization of			•••						
Chapter 2 - Review of Re-									
lated Literature									
Preparation of Letters; UP-		•							
VREB and Establish Com-									
munication for Collabora-									
tion									
Drafting and Finalization of			•	••					
Chapter 3 - Methodologies									
Proposal Document Cre-			•	•					
ation in LaTex									
Proposal Presentation				•					
Data Gathering				•••	•••				
Preprocessing					•	•••			
Drafting and Finalization of						•	••		
Chapter 4 - Results and Dis-									
cussion									
Corpus Development and							•••	•••	
Validation									
Drafting and Finalization of								••	••
Chapter 5 - Summary and									
Recommendation									
Drafting and Finalization of									•
SP Defense									
SP Defense									•

Chapter 4

Preliminary Results

This chapter presents the preliminary results of the study. This include the visual examples of the preliminary results for text and phonetic transcription.

4.1 Text and Phonetic Transcription

The figures below show the current working file of the text corpus.

English	Akeanon	IPA	Zorc IPA	SEPARATED	DAVID ZORC'S COMMENTS OR EXPLANATION
1	ako	?aˈko	?akó		
you	ikaw	?i'kaw	?ikáw		
he	imaw	?i'maw	?imáw		
vou (plural)	kamo	ka'mo	kamó		
they	sanda	san'da	sánda		
this	dava/hara	'daia/ha'ra	dáva	hará 'here it is'	Entries should be limited to a single word for sorting and practical purposes.
that	dato/hato	da'to/ha'to	dató	ható 'there it is'	
here	iva	2i'ia	2ivá		The IPA <> symbol is based on German insistence. Most languages and all Filipinos use and appreciate <y>.</y>
there	idto	?id'to	2ídto		Tagalog intonation on the final syllable has crept in here. Accent in Aklanon is on the first syllable of this word.
who	sin-o	sin'2o	sín2o		
what	ano/alin	2a'no/2a'lin	2anó	2alín	Entries should be limited to a single word for sorting and practical purposes.
where	siin	si'?in	si?ín		
when	kan-o	kan'?o	kán?o		Tagalog intonation on the final syllable has crept in here. Accent in Aklanon is on the first syllable of this word.
how	paalin	pa'?alin	pa?alín		It is my understanding that accent falls on the last syllable of this word.
not	bukon	bu'kon	bukón		
all	tanan	ta'nan	tanán		
many	abo	?a'bo?	2abó2		
some	ro iba	ro 2i'ba			ro 2ibáh literally means "the others" I Depending on context English "some" could be <may 2úna2=""></may>
few	sangkiri	sanki'ri?	sankirí?		
other	iba	?iˈba	?ibáh		It is my understanding that this word ends with a clearly pronounced [h]
one	isaea	?isa'ra	2isawá		I was wrong about the symbol for Aklanon <e> [voiced yelar semiyowel], the correct IPA symbol is <u></u></e>
two	davwa	dai wa	7.00000		
three	tatlo	tat'lo			
four	ap-at	?ap'?at			
five	lima	li'ma			
big	maeagko	mayag ko?	mawagko?	mágko?	Symbol should be <u≥ (second="" accent="" and="" is="" on="" penult="" syllable)<="" td="" the=""></u≥>
long	mahaba		maháha?		I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
wide	maeapad		mawápad		I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
thick	madamoe		madámou		I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
heavy	mabug-at		mabúg?at		I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
small	maintok		ma2intok		I do not object to the adjectives having a ma- prefix but we must insure that the ROOT WORD is also in the dictionary.
short	manaba		armton	1111011	I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
narrow	mapiot				I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
thin	manipis				I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
woman	bave		bávih		Rural folks in Aklan do not pronounce a final <> as <e> as Tagalogs do. This is another word that ends in -h.</e>
man (adult	eaki		wáki		and the state of t
man	tawo		ujuki		
child	unga				
wife	asawa				

Figure 4.1: Preliminary transcription of Akeanon words into text along with their phonetic representations.

spelling	morphology	IPA	source	notes	dialect	etymo	form	ctr	# vul s	sig	pos	domain definition
	abaníko					Sp	†abanicar				RV4	to fan
abáno(h)			Sp								n	large cigar
abánti(h)			Sp								RV6	to go ahead, advance, go forward
	abánti										RV3	to endure, put up with
ábang											RV3	to obstruct, block, hinder; waylay, ambush
ábangay											n	[fernwith broad pods]
abát											RV1	without invitation)
abáw											intj	wow, gosh, oh (expression of amazement)
ábay			Tag								n	sponsor (at wedding or baptism)
ábi											Dp	Particle used in excuse or explanations; "but" "well" "because"
ábi(h)											Dp	"Gimme" "Come on." [requesting particle]
ábi-ábi											RV3	to welcome, entertain
	maabí-abíhon										adj	hospitable, friendly
	ábyan										n	friend
abíl									Υ		adj	half-castrated, having only one testicle
abenturár			Sp								RV6	to adventure, go onan adventure
	abenturera										n	adventurer [female], adventuress
	abenturero										n	adventurer [male]
abiríya			Sp								n	to gel stalled, break down [car]
abíso(h)			Sp								n	note, notice
	abíso										RV3	to notify, inform
abnórmal			Eng								adj	abnormal, out of the ordinary
abó			Oak								n	great grandfather
abó(h)											RV1	ash(es),; to make ashes
	abuhón										adj	grey, gray, ashen color(ed)
	kaabohan										n	a place totally covered with ashes
abó											adj	much, many, a lot of [refering to things which canbe counted or enumerated]
	abó										RV1	to increase, make/become greater [in supply or number]
kaabo-án											n	most; the majority
makiabó											adj	fond of plenty, used to a lot
pakaabó											CV	to consider as plenty [even if only a few]

Figure 4.2: A keanon Dictionary transcription.

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Appendix A

Research Ethics Documents

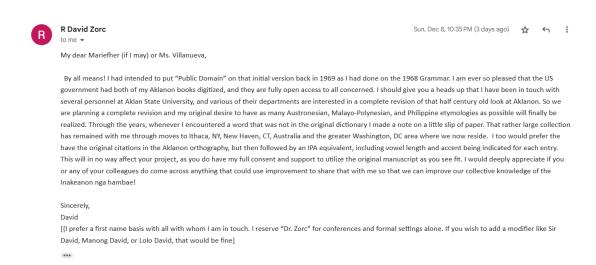


Figure A.1: Screenshot of email from Dr. Zorc granting permission to use the Aklanon-to-English Dictionary.

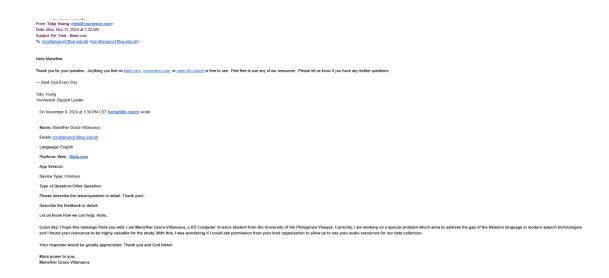


Figure A.2: Screenshot of email from Bible.com granting permission to use audio resources and transcription from their website.

Informed Consent

om the University of the Philippines ergraduate research for our special in Speech and Text Corpus Towards
reciated. We would like to extend to of our study. As a native speaker of as in developing an Akeanon speech oluntary. If you agree to participate, throughout the duration of the study be taken against you.
O Akeanon words, provided by the be used for the purpose of this study, ks related to the study. Furthermore, sure anonymity.
r information sheet attached to this
onsent
above. I've had the chance to inquire actory response. I consent voluntarily

Figure A.3: Informed consent to be given to the participants of the study.

Information Sheet

About the Researchers. This special problem is undertaken by Joshua Villanueva and Mariefher Grace Villanueva, fourth-year students from the University of the Philippines Visayas, under the supervision of Dr. Francis D. Dimzon (Assistant Professor for Computer Science), as a requirement towards a bachelor's degree in computer science.

About the Project. This special problem aims to develop a comprehensive text and speech corpus and build a model as a foundation for an automatic speech recognition (ASR) system for standardized Akeanon language. As part of the data collection, the researchers must gather voice recordings from native speakers of the language, speaking a collection of Akeanon words.

Participant Selection and How to Participate in the Study. You are invited to participate in the study because you belonged to the inclusion criteria listed above. To participate, you will be voice-recorded by the researchers while speaking a provided set of Akeanon words. As a way of compensation for participating in the study, you will receive snacks during your session.

Data Management. The voice recordings will solely be used for research purposes, and any authorized use by the researchers. The researchers, supervisor, and possible collaborators will have access to the recordings. Rest assured that access to these recordings is highly restricted, and they will not be available to the public. Though the results of the study may be used for academic publication but rest assured that your anonymity is maintained.

Your Rights as a Participant. During your session, you have the right to stop your participation and withdraw from the study, at any stage of the recording.

For Questions or Comments. Should you have any questions or feedback regarding the study, you can contact:

Mariefher Grace Villanueva	Joshua Villanueva	Dr. Francis D. Dimzon
Primary Researcher	Primary Researcher	Thesis Adviser
Division of Physical Sciences and	Division of Physical Sciences	Division of Physical Sciences
Mathematics	and Mathematics	and Mathematics
College of Arts and Science	College of Arts and Science	College of Arts and Science
University of the Philippines	University of the Philippines	University of the Philippines
Visayas	Visayas	Visayas
mzvillanueva l@up.edu.ph	jcvillanueva5@up.edu.ph	fddimzon1@up.edu.ph
09273182739	09944616691	

Research Ethics Board Approval. This research will be reviewed by the University of the Philippines Visayas Research Ethics Board. If you have any concerns about the conduct of the research, please contact the Office of the Vice Chancellor for Research and Extension through overe.upvisayas@up.edu.ph.

Figure A.4: Information sheet attached to the informed consent.

Appendix B

Resource Persons

Ms. Hazel Anne Cipriano

Linguist

University of the Philippines Diliman

havcipriano@gmail.com

Dr. John Orbista

Local Collaborator

College of Teacher Education

Aklan State University

johnorbista@gmail.com

Dr. R. David Zorc (Lolo David)

Linguist

Language Research Center, Hyattsville, MD - retired

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Appendix C

Initial Results of the Text Corpus

English	Akeanon	IPA	Zorc IPA	SEPARATED	DAVID ZORC'S COMMENTS OR EXPLANATION
1	ako	?aˈko	?akó		
you	ikaw	?i'kaw	?ikáw		
he	imaw	?i'maw	2imáw		
ou (plural)	kamo	ka'mo	kamó		
thev	sanda	san'da	sánda		
this	daya/hara	'daja/ha'ra	dáya	hará 'here it is'	Entries should be limited to a single word for sorting and practical purposes.
that	dato/hato	da'to/ha'to	dató	ható 'there it is'	
here	iva	?i'ia	?ivá		The IPA <i> symbol is based on German insistence. Most languages and all Filipinos use and appreciate <y>.</y></i>
there	idto	?id'to	?ídto		Tagalog intonation on the final syllable has crept in here. Accent in Aklanon is on the first syllable of this word.
who	sin-o	sin'?o	sín?o		
what	ano/alin	?a'no/?a'lin	?anó	?alín	Entries should be limited to a single word for sorting and practical purposes.
where	siin	si'?in	si?ín		
when	kan-o	kan'?o	kán?o		Tagalog intonation on the final syllable has crept in here. Accent in Aklanon is on the first syllable of this word.
how	paalin	pa'?alin	pa?alín		It is my understanding that accent falls on the last syllable of this word.
not	bukon	bu'kon	bukón		
all	tanan	ta'nan	tanán		
many	abo	?a'bo?	?abó?		
some	ro iba	ro ?i'ba			ro ?ibáh literally means "the others" Depending on context English "some" could be <may ?úna?=""></may>
few	sangkiri	sanki'ri?	sankirí?		
other	iba	?i'ba	?ibáh		It is my understanding that this word ends with a clearly pronounced [h]
one	isaea	?isaˈra	?isawá		I was wrong about the symbol for Aklanon <e> [voiced velar semivowel], the correct IPA symbol is <u></u></e>
two	daywa	dai wa			
three	tatlo	tat'lo			
four	ap-at	?ap'?at			
five	lima	li'ma			
bia	maeagko	mayag'ko?	mawagko?	wágko?	Symbol should be <u> and accent is on the penult (second syllable)</u>
long	mahaba		mahába?	hába?	I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
wide	maeapad		mawapad		I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
thick	madamoe		madámou	dámou	I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
heavy	mabug-at		mabúg?at		I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
small	maintok		ma?íntok	2íntok	do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
short	manaba				I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
narrow	mapiot				I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
thin	manipis				I do not object to the adjectives having a ma- prefix, but we must insure that the ROOT WORD is also in the dictionary.
woman	baye		báyih		Rural folks in Aklan do not pronounce a final <i> as <e> as Tagalogs do. This is another word that ends in -h.</e></i>
nan (adult	eaki		wáki		
man	tawo				
child	unga				
wife	asawa				

Figure C.1: Partial results of transcription for the Swadesh 207 Word List.

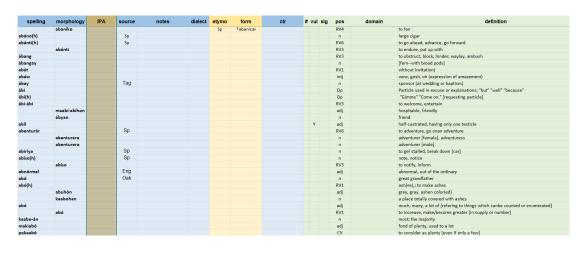


Figure C.2: Partial results of transcription of the Aklanon-to-English Dictionary.