

# Applying Swin Architecture to Diverse Sign Language Datasets

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#### **ABSTRACT**

In the era of AI, understanding non-verbal communication is paramount. This research utilizes Al Swin Architecture to bridge communication gaps in sign language communities, comparing classification of American (ASL) and Taiwanese Sign Languages (TSL).

### INTRODUCTION

Incorporating ASL into digital platforms enhances accessibility. This project focuses on a website with ASL features, including a fingerspelling dictionary and a practice section. Unique to this study is the development of a custom ASL dataset. The practice section employs the Swin Architecture for data classification. Leveraging Al and Deep Learning technologies, particularly the Hierarchical Vision Transformer using Shifted Windows (Swin) model, the project achieves 100% accuracy in ASL dataset classification. Custom datasets of ASL and TSL including both alphabet letters and short phrases (Fig. 1.) were created especially for this project. The prototype of ASL Educational App was then developed (Fig.2.) and currently being tested.

### **MATERIALS** and **METHODS**

This research utilizes the Swin Transformer model, introduced by Liu et al. from Microsoft Research [1, 2, 3], for sign language recognition, focusing on ASL and TSL. The methodology is structured into five main phases: describing diverse datasets, used for the study, understanding of Swin Architecture, model training, results evaluation, and app development.

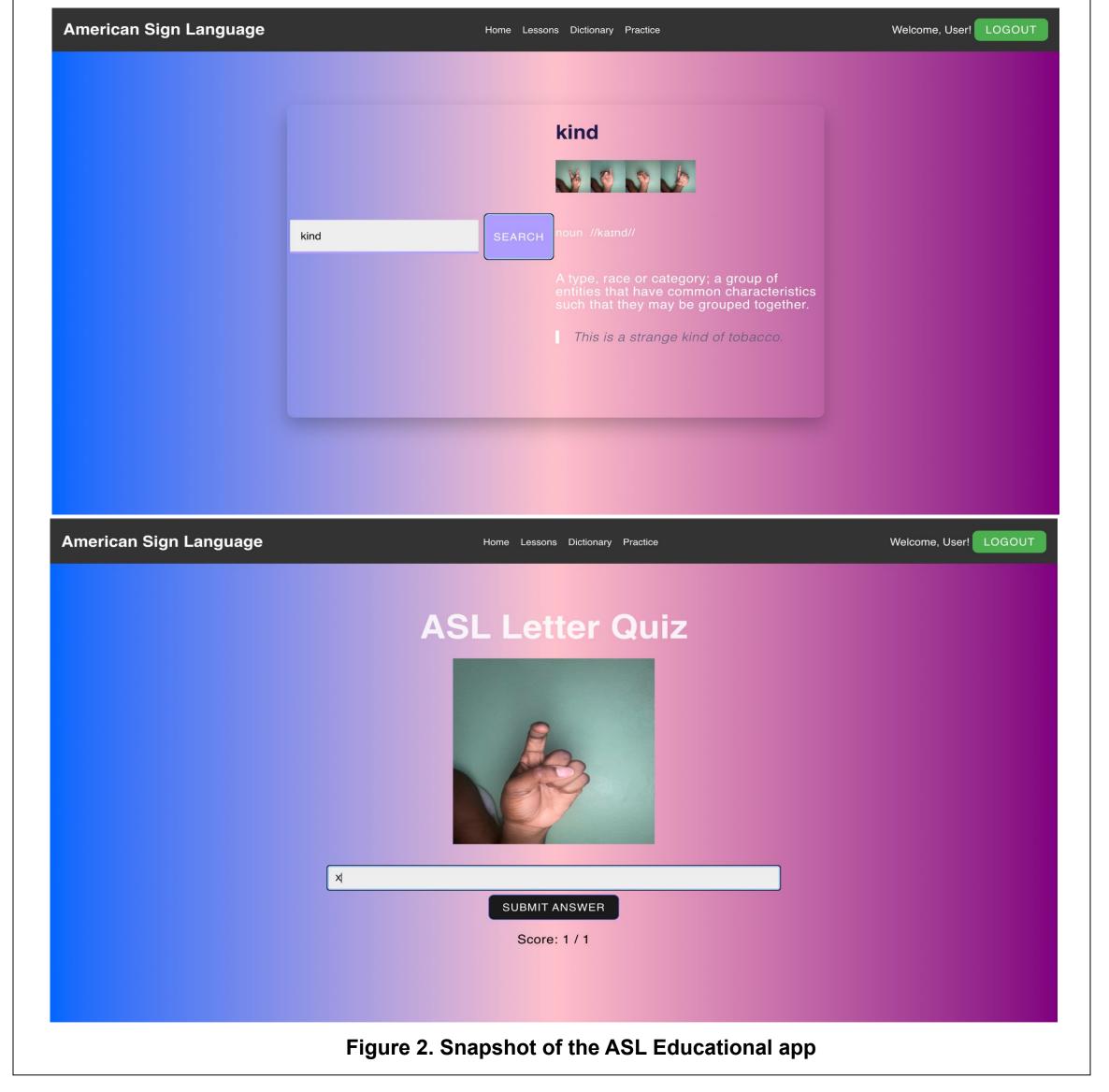
Each phase is crucial for the Sign Language electronics recognition and translation.

# **Project Datasets** iloveyou A\_hello iloveyou A\_hello A\_than

Figure 1. Snapshot of the project dataset: Alphabet Letters (left) vs Short phrases (right) in ASL and TSL

## **ASL Educational App**

Features on the website utilize an ASL dataset for fingerspelling, incorporating elements such as a dictionary and quiz matching.



### **Current Result and Future Work**

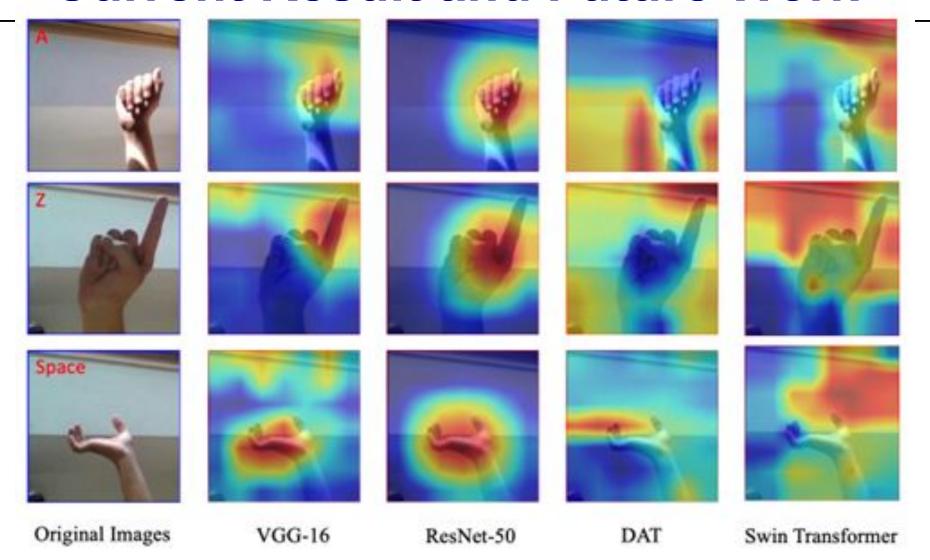


Figure 3. The CAMs of the Swin Transformer in comparison with other DL models

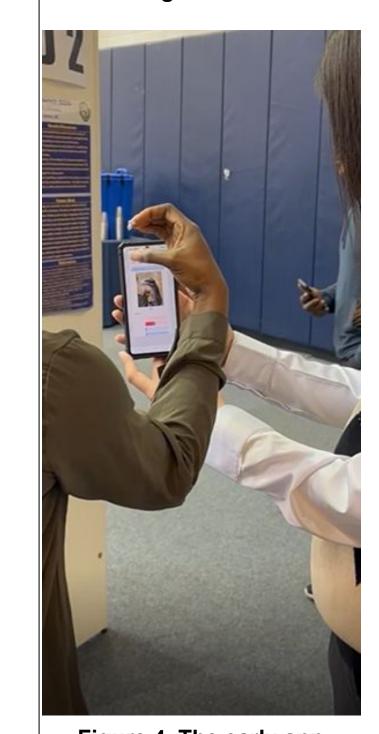


Figure 4. The early app prototype under test

Various AI models were trained on ASL TSL datasets, and achieving up to 100% of accuracy. Future work includes exploring other sign languages, refining the website's features and developing our own bias mitigation strategies for Transformers like Swin. The associated under paper consideration by Electronics Q2 journal [4] of the and goal researchers is to publish current results of this work and continue it further in 2024-2025.

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# References

