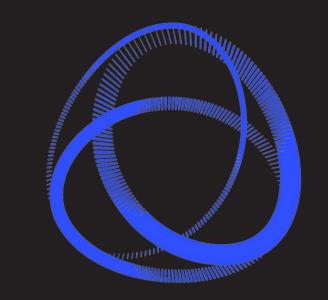
Al-Based ASL Fingerspelling Recognition Using the Google Kaggle Dataset for Automated Kiosk Transactions

A Literature Review and Project Application

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Introduction

- ► According to Ethnologue (2023); Mitchell et al. (2006), there are at least 1 million if not more American Sign Language (ASL) users in the United States alone who are a part of the Deaf and Hard of Hearing (DHH) community.
- Fingerspelling is a critical component of ASL and other sign languages, and is used to spell out words, names, and proper nouns that lack direct signs.
- ► Current recognition solutions for ASL, don't include fingerspelling, and typically don't work effectively in the real world.
- ► Legal concerns from the largest and most used ChicagoWild/+ dataset, have made it difficult for commercial application.
- ➤ A new dataset released on Kaggle by Google (Manfred Georg et al., 2023) has provided a new opportunity to explore how applying Al could improve, assist, and provide options to the lives of the DHH and the wider community?



Methodology

IEEE Xplore® Digital Library

Literature Identification

Association for Computing Machinery

ScienceDirect[®] Google Scholar

Literature Evaluation

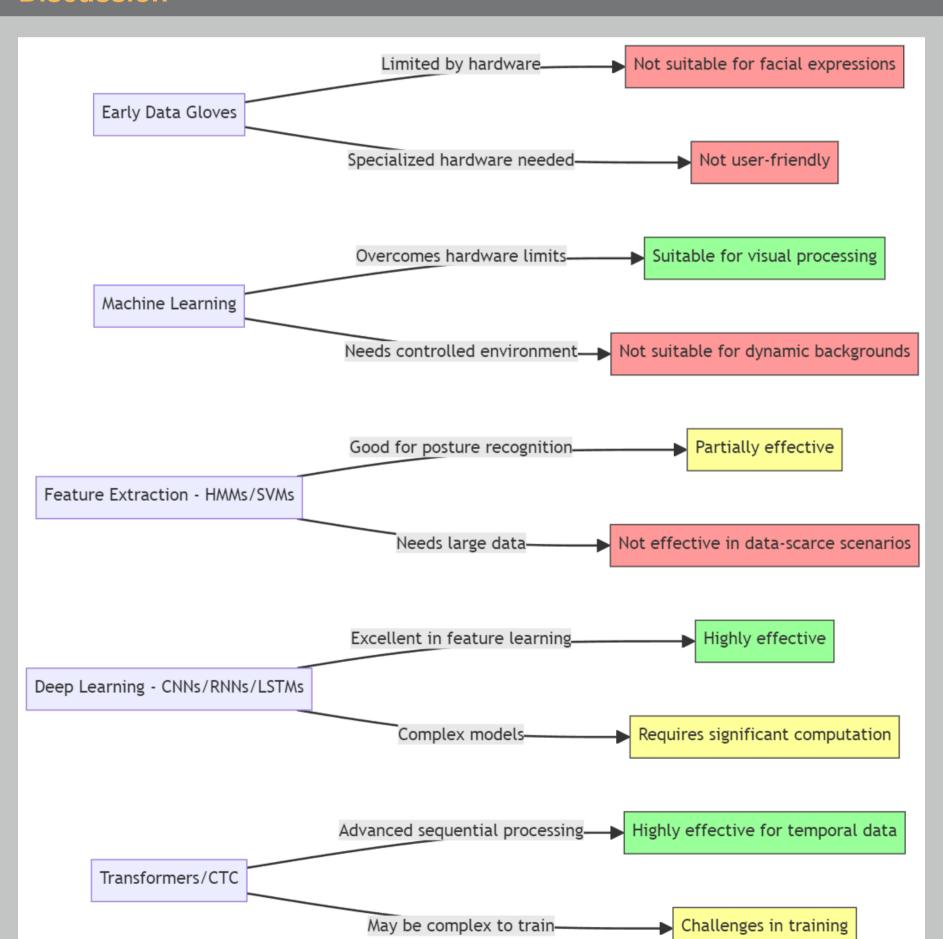
- Conducted research across four academic databases.
- Focused on papers published within the last 5 years (2018-2023).
 Limited to peer-reviewed journal
- 3. Limited to peer-reviewed journal articles, conference papers, and high-quality theses.
- 4. Search conducted in English only.
- 5. Search period: November to December 2023.
- 6. Search terms such as "ASL fingerspelling", "ASL recognition in real-time", "Deep learning for ASL fingerspelling".

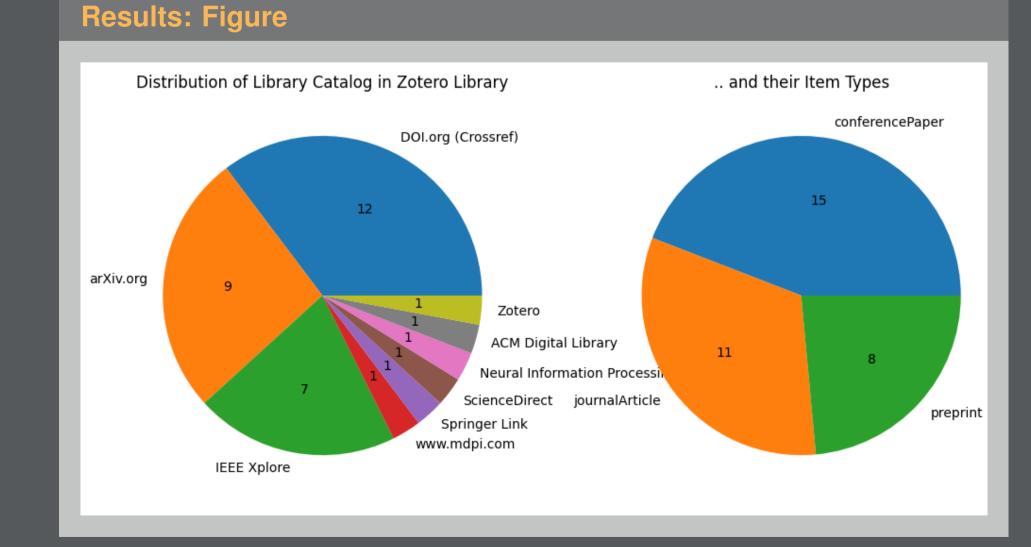
- Literature Evaluation
- Abstract → relevance → full reading of selected papers
- 2. Summarization and analysis of findings in a table
- 3. Inclusion Relevance to research
- 4. Specific focus on ASL fingerspelling
- 5. Use of machine learning (ML) models in sign language interpretation
- 6. Utilization of recognized datasets relevant to ASL recognition
- 7. Clear methodology, defined objectives, data analysis
- 8. High citation counts preference
- Excluded Editorials, opinion pieces, and non-peer-reviewed articles

Results: Table

Model	Key Insights and Performance	Citation	
RNN, LSTM, Attention	Recognition and translation of ASL glosses; GRR: 86%, GER: 23%. Challenges in real-time recognition.	S Kuma et al. (2018)	
CNN, SSD, FCN	High accuracy in vision-based translation; Accuracy: 92.21%. Robustness in ASL recognition.	Abiyev et al (2020)	
Transformers, CTC	State-of-the-art results in ASL recognition; WER, BLEU-4 scores. Translation challenges addressed.	Cihan Cam- goz et al (2020)	
ResNet, Bi-LSTM	Recognition using optical flow; Letter accuracy: 57%. Focus on 'wild' conditions and occlusions.	Kabade et al. (2023)	
2D/3D-CNN, Bi- LSTM	Superior detection in diverse environments; AP@IoU: 0.495, MSA: 0.386. Fine-grained handshapes analysis.	Shi et al (2021)	
Fine-Grained Visual Attention	Improved recognition with Transformer model; Letter Accuracy: 46.96% (dev). Addressing video data challenges.	Gajurel et al (2021)	
Extremely Condensed Summary of ASL Fingerspelling Recognition Models (2018-2023)			

Discussion





References

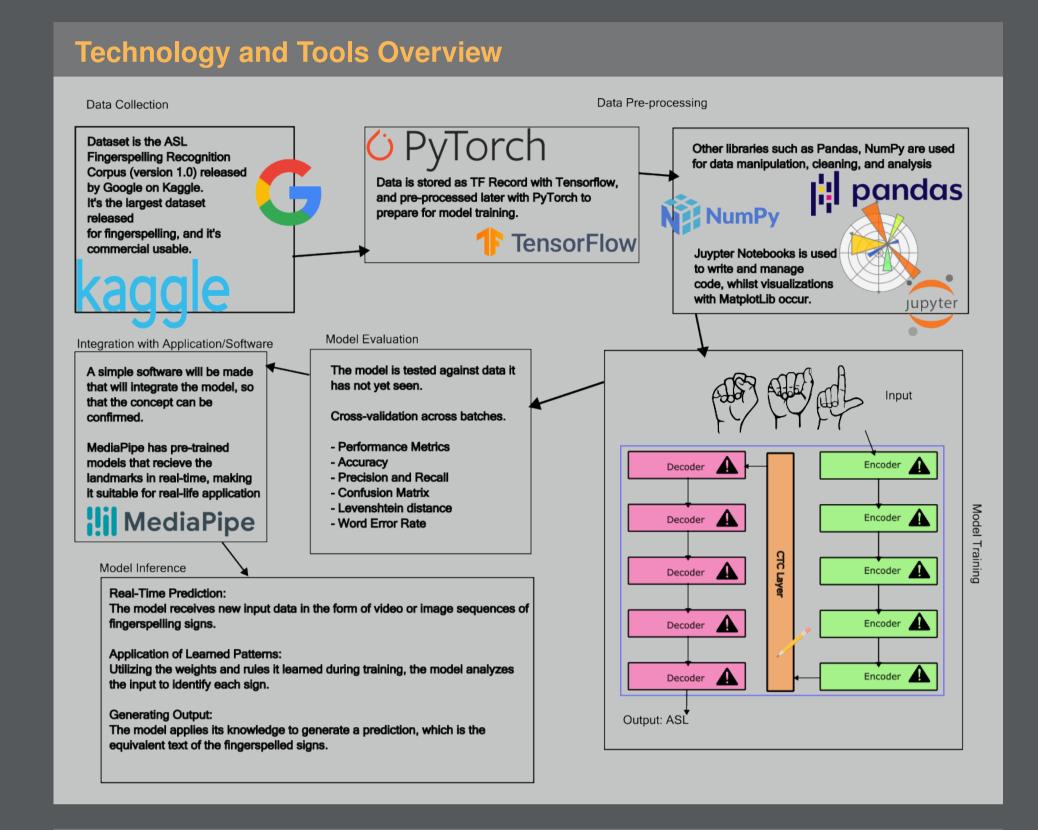
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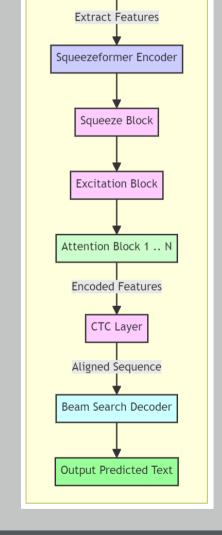
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Conclusions

- CNNs, RNNs, LSTMs, and Transformers have shown promise in recognizing ASL fingerspelling.
- ► Real-world challenges, including rapid signing, varied hand orientations, and diverse environmental conditions, pose substantial difficulties for current models.
- ► Techniques such as data augmentation, transfer learning, and the use of pre-trained models are critical in overcoming these obstacles.
- ➤ State-of-the-art methods are evolving, with recent developments in Transformers and Squeezeformer models, and real-world applications are expanding, notably in kiosk systems.

Model: A combination of the state-of-the-art technologies would look like



Input Sequence

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