# **American Sign Language Alphabet Classification**

## **G07**

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#### **Project Overview**

By harnessing deep learning models and a combination of supervised and semi-supervised decision tree models, our project aims to:

- Develop a robust ASL alphabet recognition tool.
- Comparing models on various metrics.
- Achieve higher accuracy models.

#### **Project Goal**

Our project focuses on developing a robust ASL alphabet recognition tool that achieves higher accuracy in interpreting American Sign Language (ASL) alphabets.

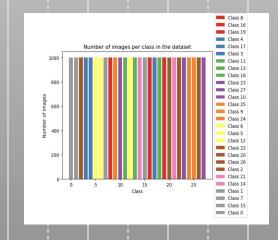
This allows us to overcome limitations like:

- Reduce communication barriers.
- Promote inclusivity.
- Enable seamless communication for individuals with and without hearing impairments.

#### **Dataset:**

- American Sign language (Link: <u>American Sign Language | Kaggle</u>)
- Total number of Images in the dataset are 165K (Size: 4.98 GB)
- Total Categories '28'

Dataset attributes	ASL dataset	
Before Trimming Dataset		
Image size	400x400x3	
Image count	165k	
Class count	28	
After Trimming Dataset		
Image size	256x256x3	
Class count	28	
Image count	28,000	

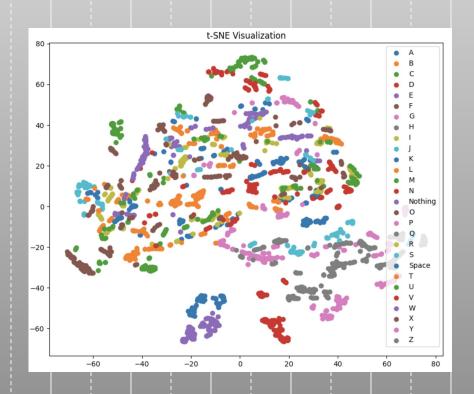


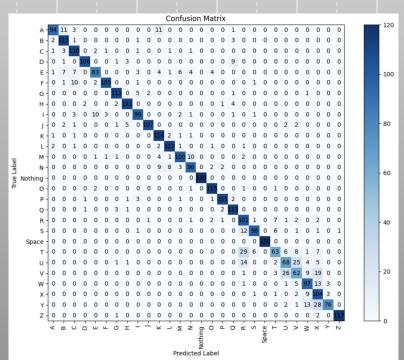
#### Methodology

1. Supervised learning Classification with Decision Trees

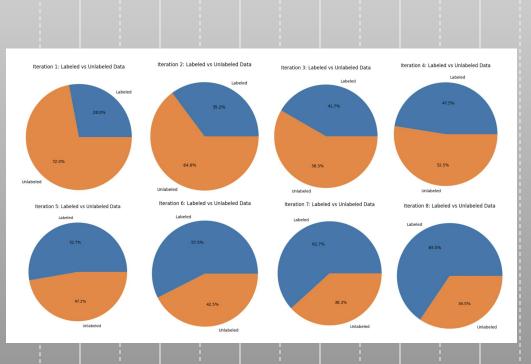
- 2. Semi-supervised learning Classification with Decision Trees
- 3. Supervised learning Classification with CNN model.

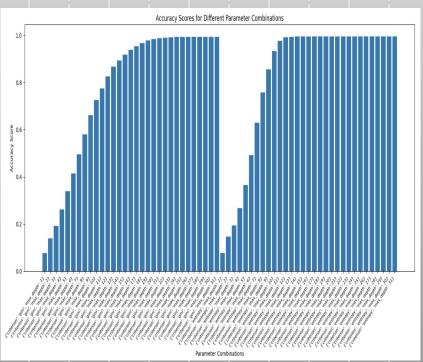
### Results for Supervised learning Classification with Decision Trees





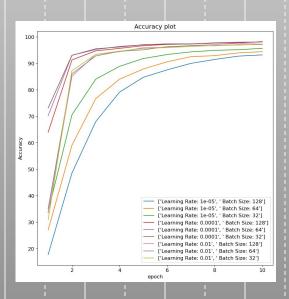
#### Results for Semi-supervised learning Classification with Decision Trees

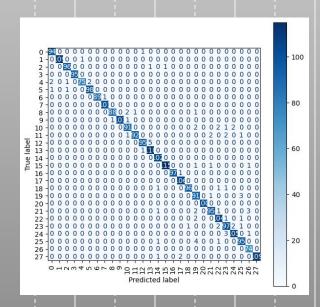




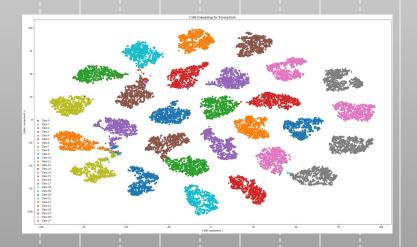
#### Results for Supervised learning Classification with CNN model

- Achieved test accuracy of 98.5% (Batch size: 128 and learning rate: 0.0001)
- CNN model gives the best result on the dataset as compared to decision trees.
- Residual connections enable deeper networks to be trained more effectively.
- Hyper parameter tuning with batch size (32, 64, 128) and learning rate (0.01, 0.0001, 0.00001).





- Larger batch size less training time.
- 0.0001 learning rate lead to most optimal accuracies.



Metrics	Accuracy	Precision	Recall	F1-measure
CNN	0.99	0.99	0.99	0.99
Supervised	0.89	0.88	0.87	0.87
Semi-Supervised	0.85	0.84	0.82	0.82

t-SNE Plot

**Metrics Evaluation** 

#### References

- <a href="https://www.kaggle.com/datasets/kapillondhe/american-sign-language">https://www.kaggle.com/datasets/kapillondhe/american-sign-language</a>
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- https://www.projectpro.io/recipes/optimize-hyper-parameters-of-decisiontree-model-using-grid-search-in-python
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